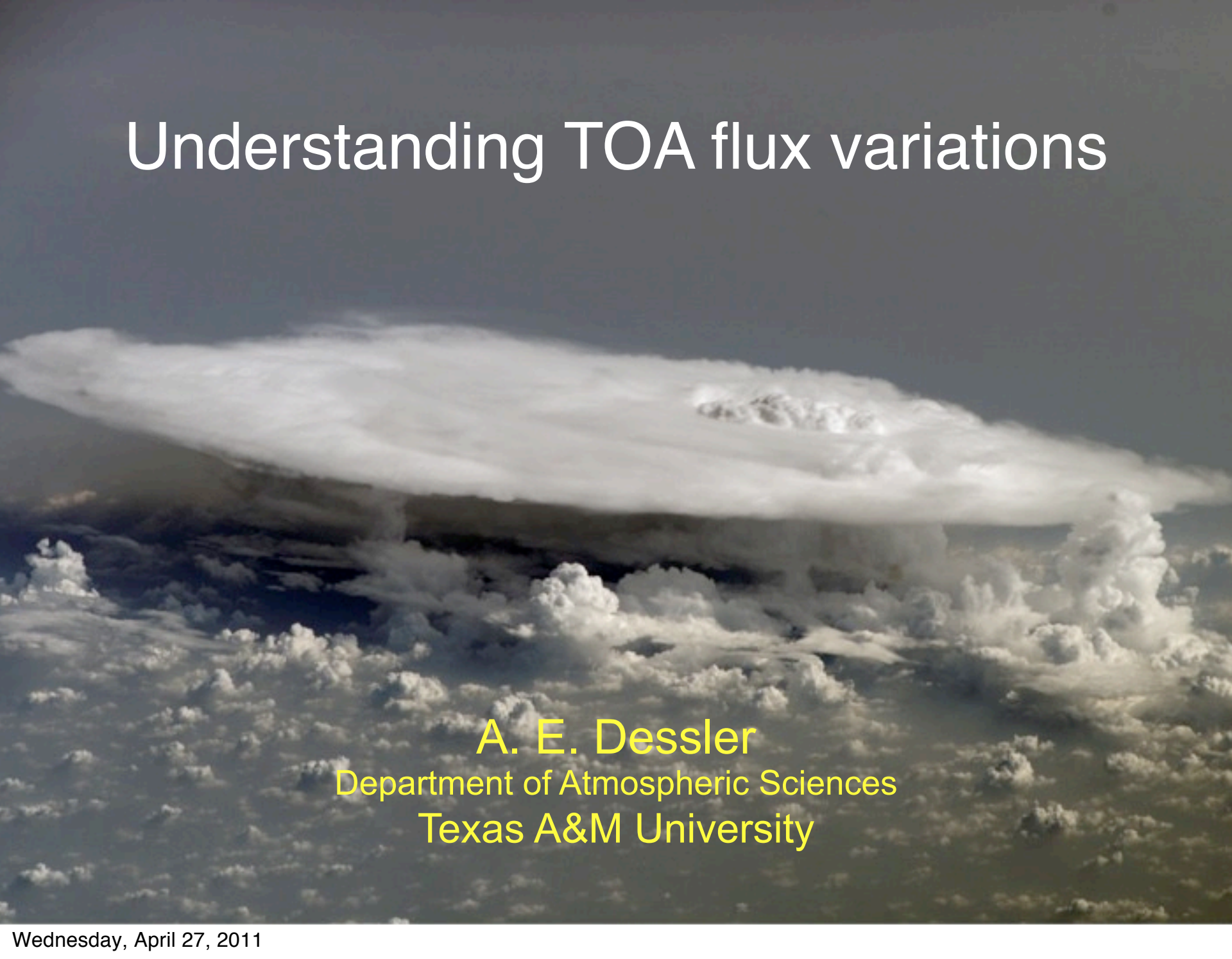
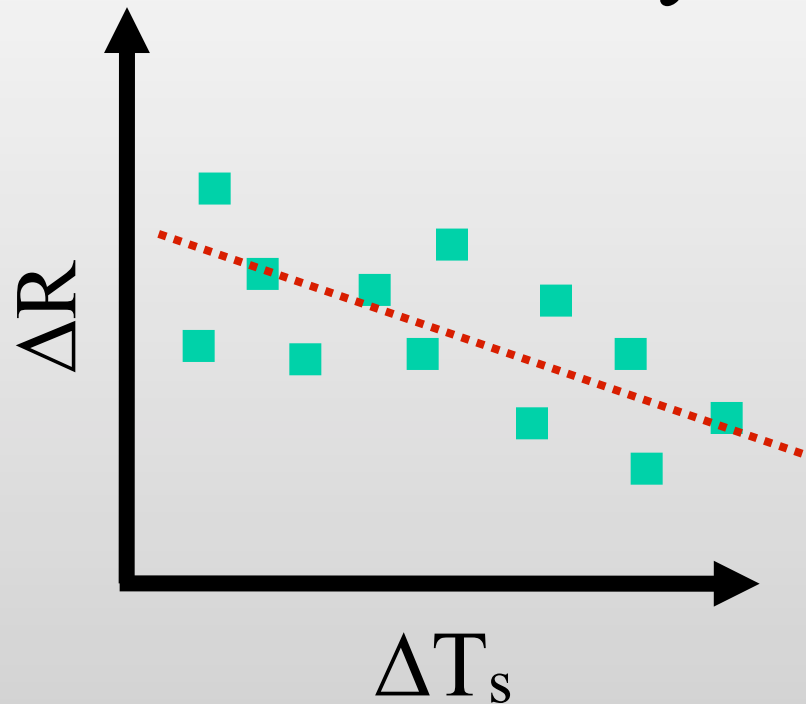


# Understanding TOA flux variations

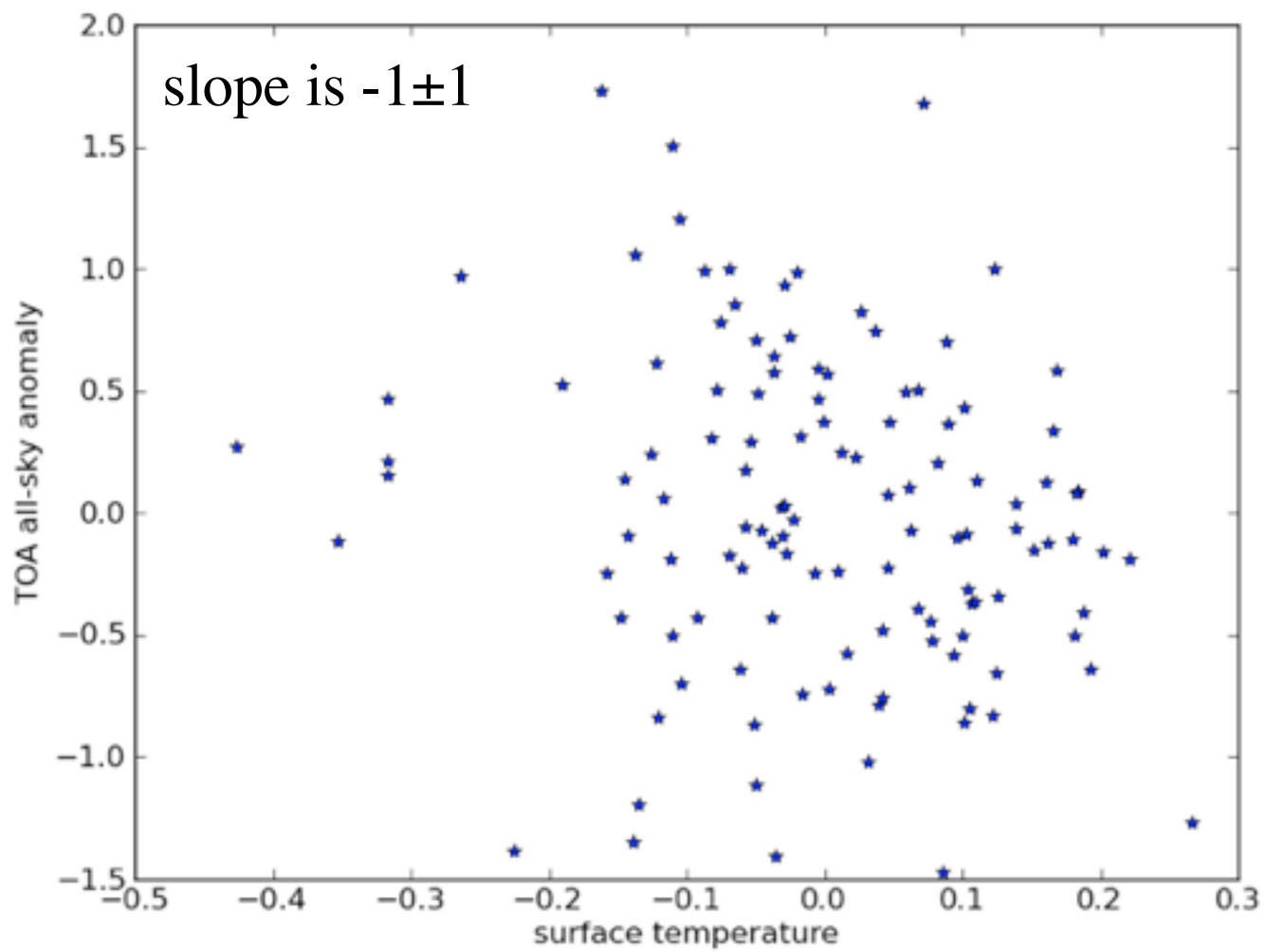


A. E. Dessler  
Department of Atmospheric Sciences  
Texas A&M University

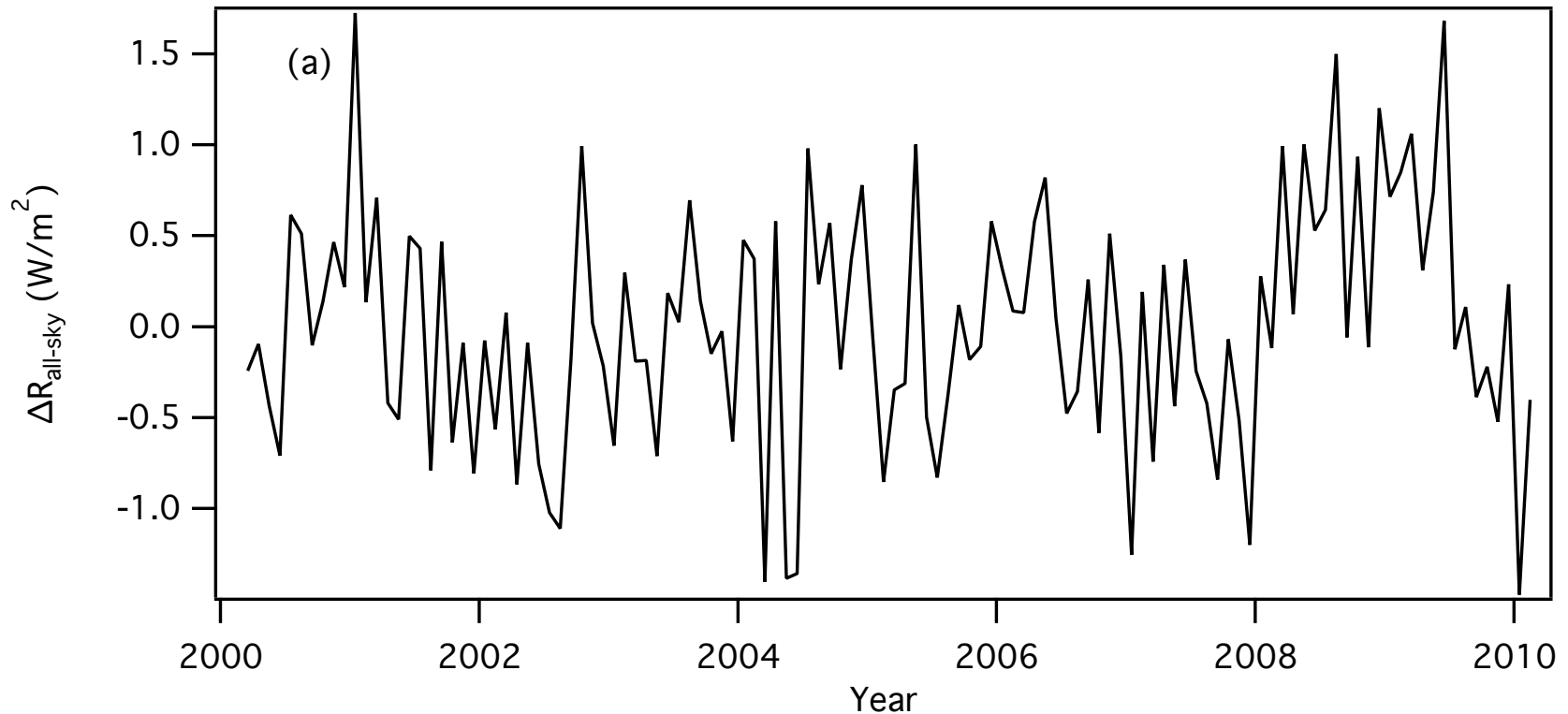
... to better understand climate  
sensitivity



e.g., Forster and Gregory, 2006;  
Lin et al., JQSRT, 2010; Murphy, 2010  
all fluxes in this analysis are downward positive



# CERES top-of-atmosphere (TOA) net flux SSF, 1-deg monthly avg., Ed. 2.5



all fluxes in this analysis are downward positive

$$\Delta R_q = \sum_{x,y,z} \frac{\partial R}{\partial q(x,y,z)} \Delta q(x,y,z)$$

$$\Delta R_q = \sum_{x,y,z} \frac{\partial R}{\partial q(x,y,z)} \Delta q(x,y,z)$$

water vapor anomaly



$$\Delta R_q = \sum_{x,y,z} \boxed{\frac{\partial R}{\partial q(x,y,z)}} \Delta q(x,y,z)$$

water vapor anomaly

Use pre-computed kernels from Soden et al., 2008, see also Shell et al. [2008]

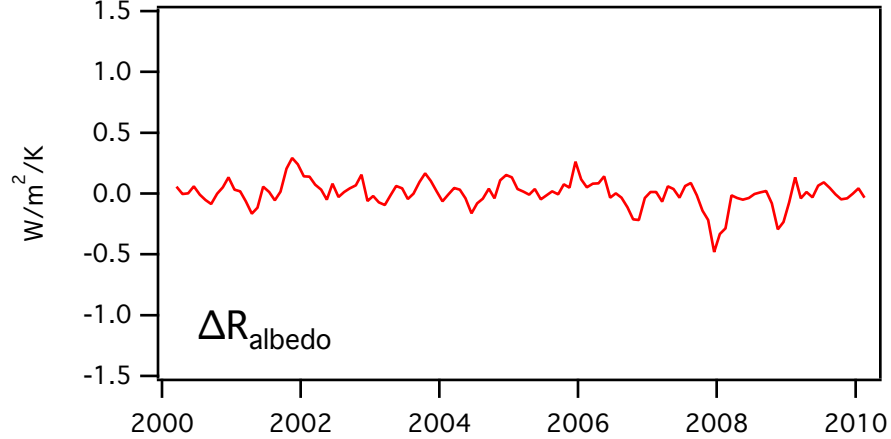
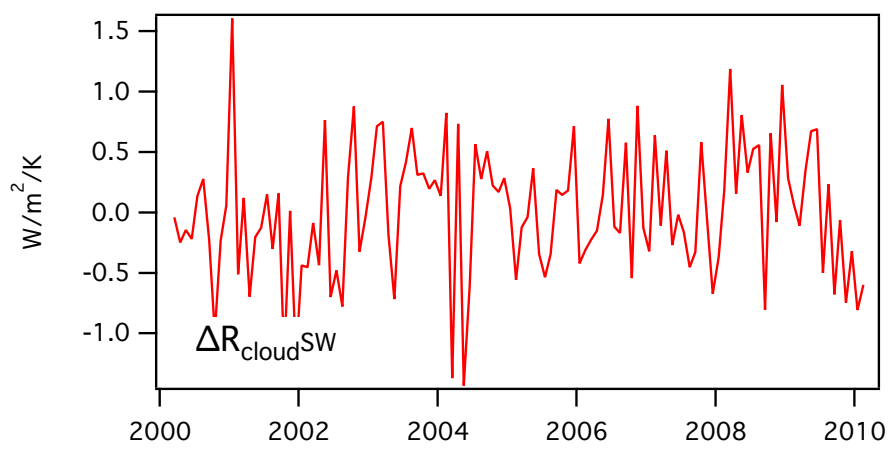
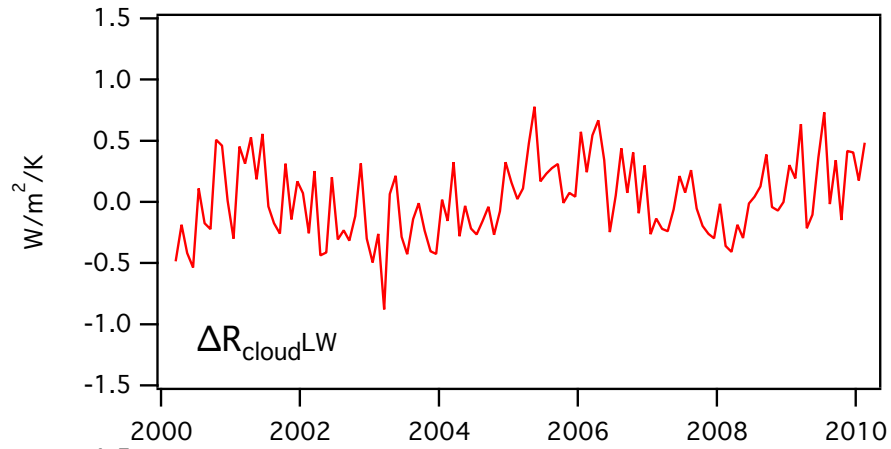
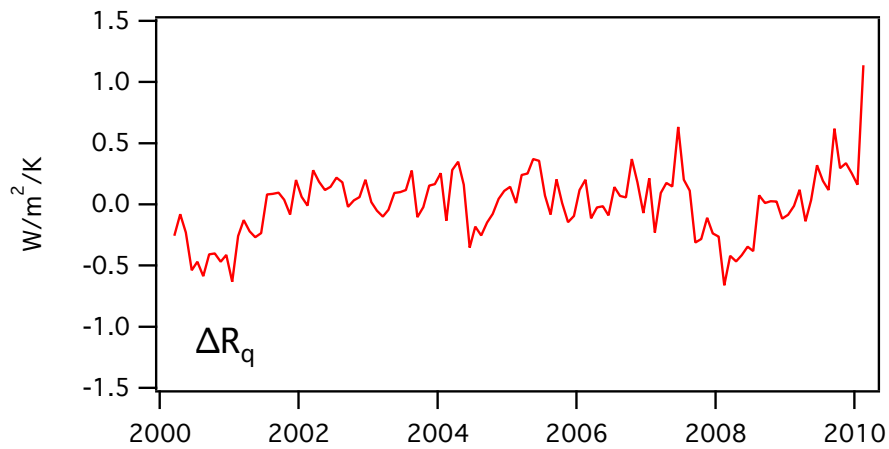
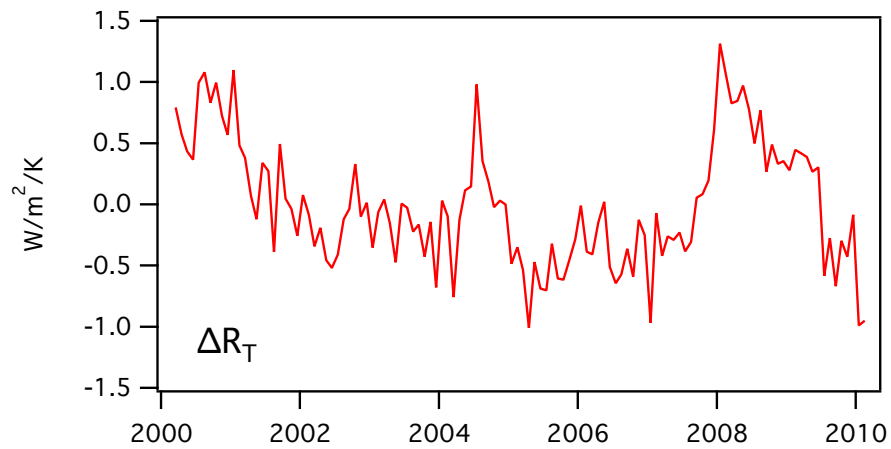
$$\Delta R_q = \sum_{x,y,z} \boxed{\frac{\partial R}{\partial q(x,y,z)}} \Delta q(x,y,z)$$

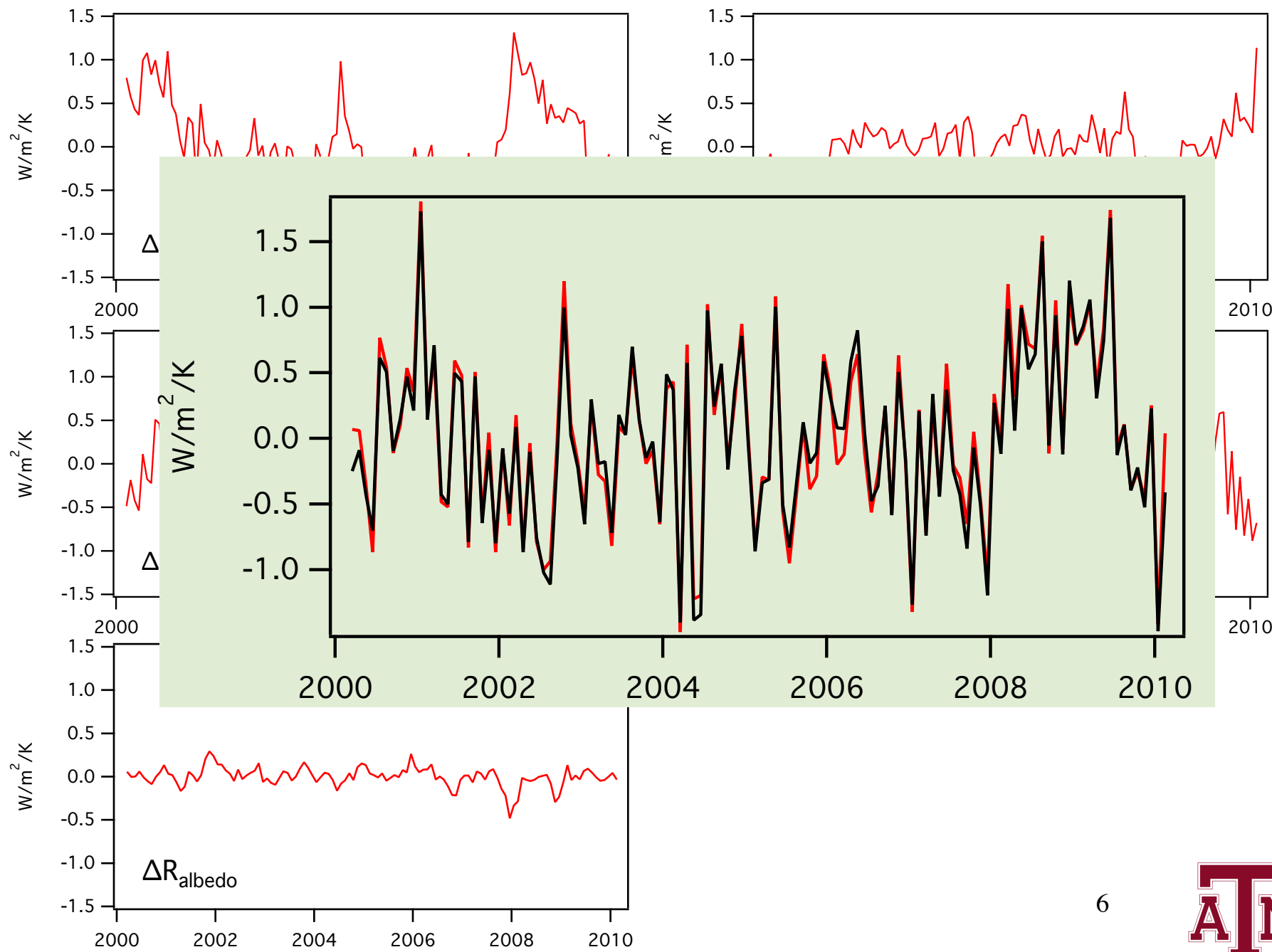
water vapor anomaly

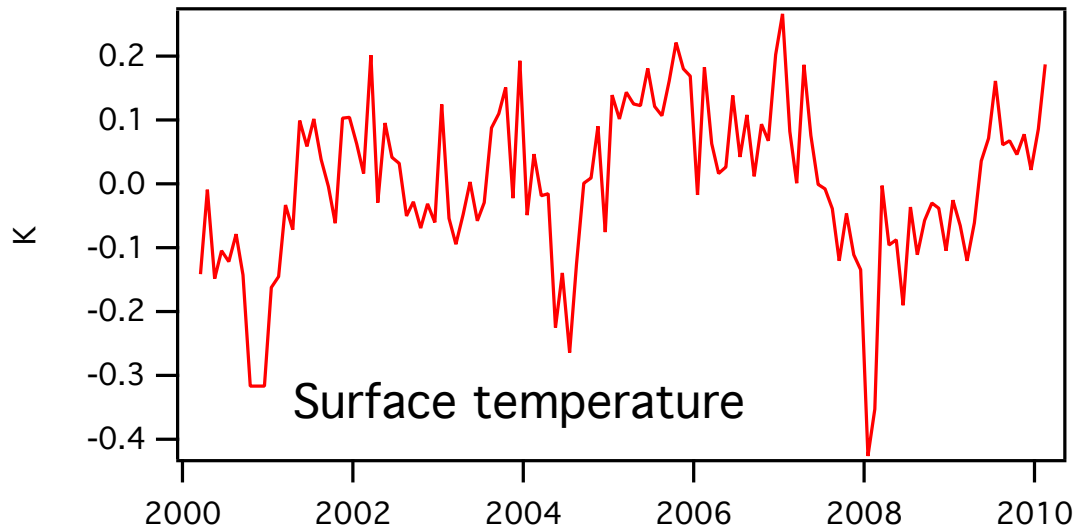
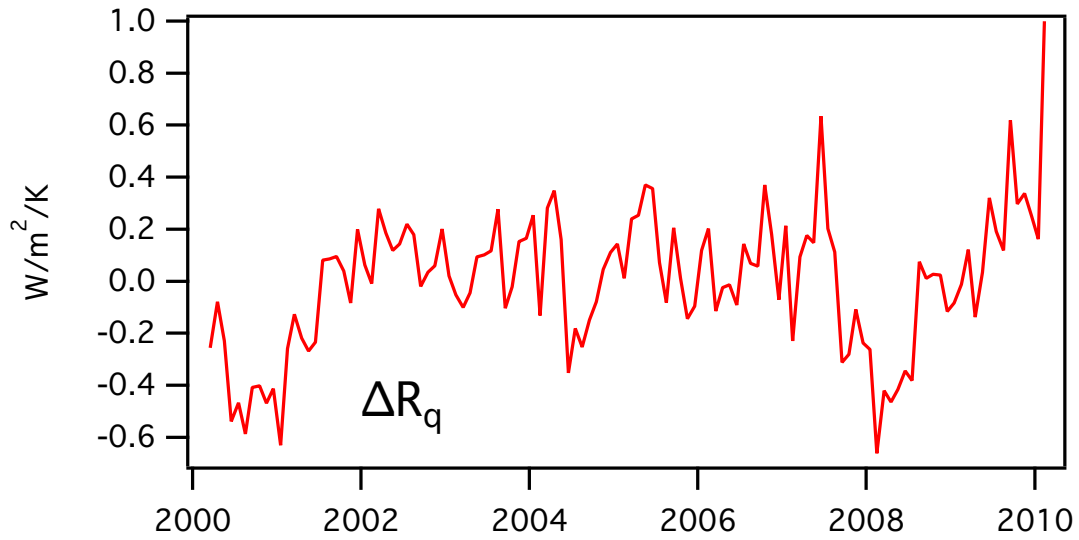
Use pre-computed kernels from Soden et al., 2008, see also Shell et al. [2008]

$$\Delta R_T = \sum_{x,y,z} \frac{\partial R}{\partial T(x,y,z)} \Delta T(x,y,z)$$

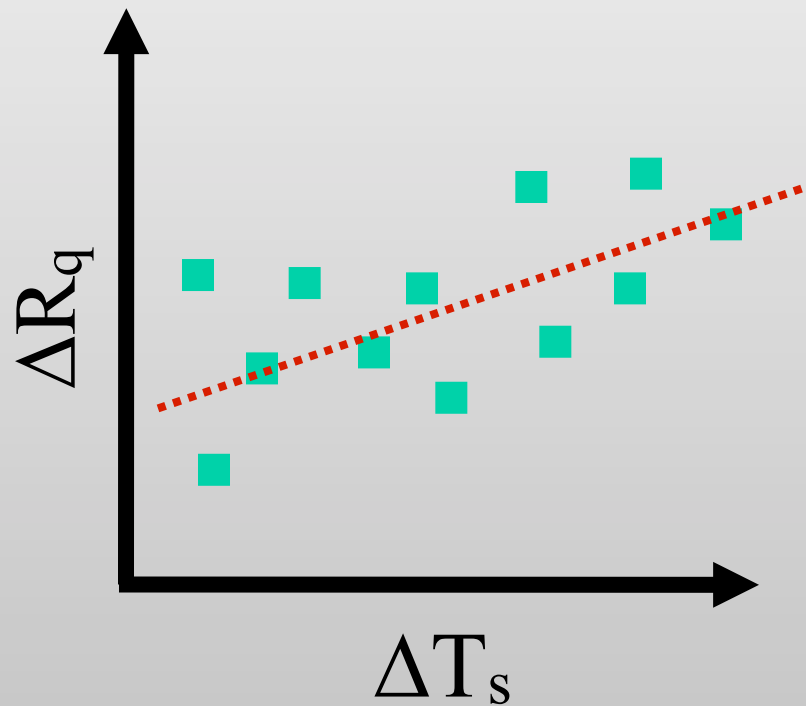


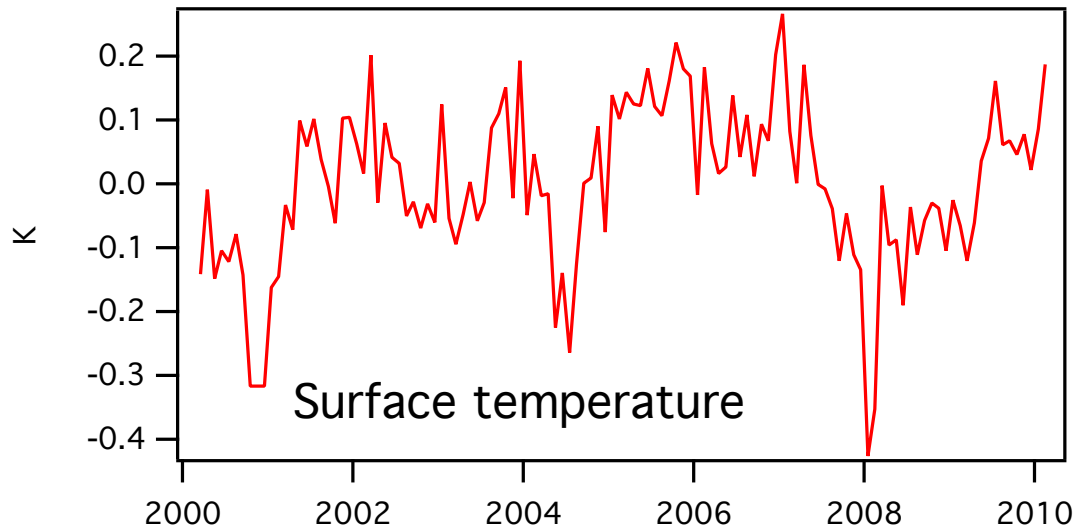
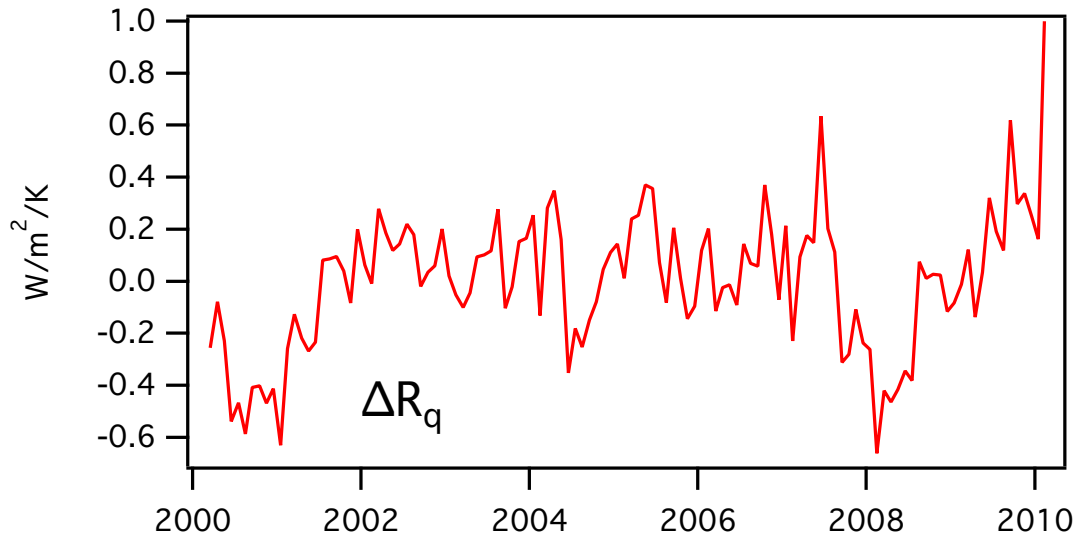






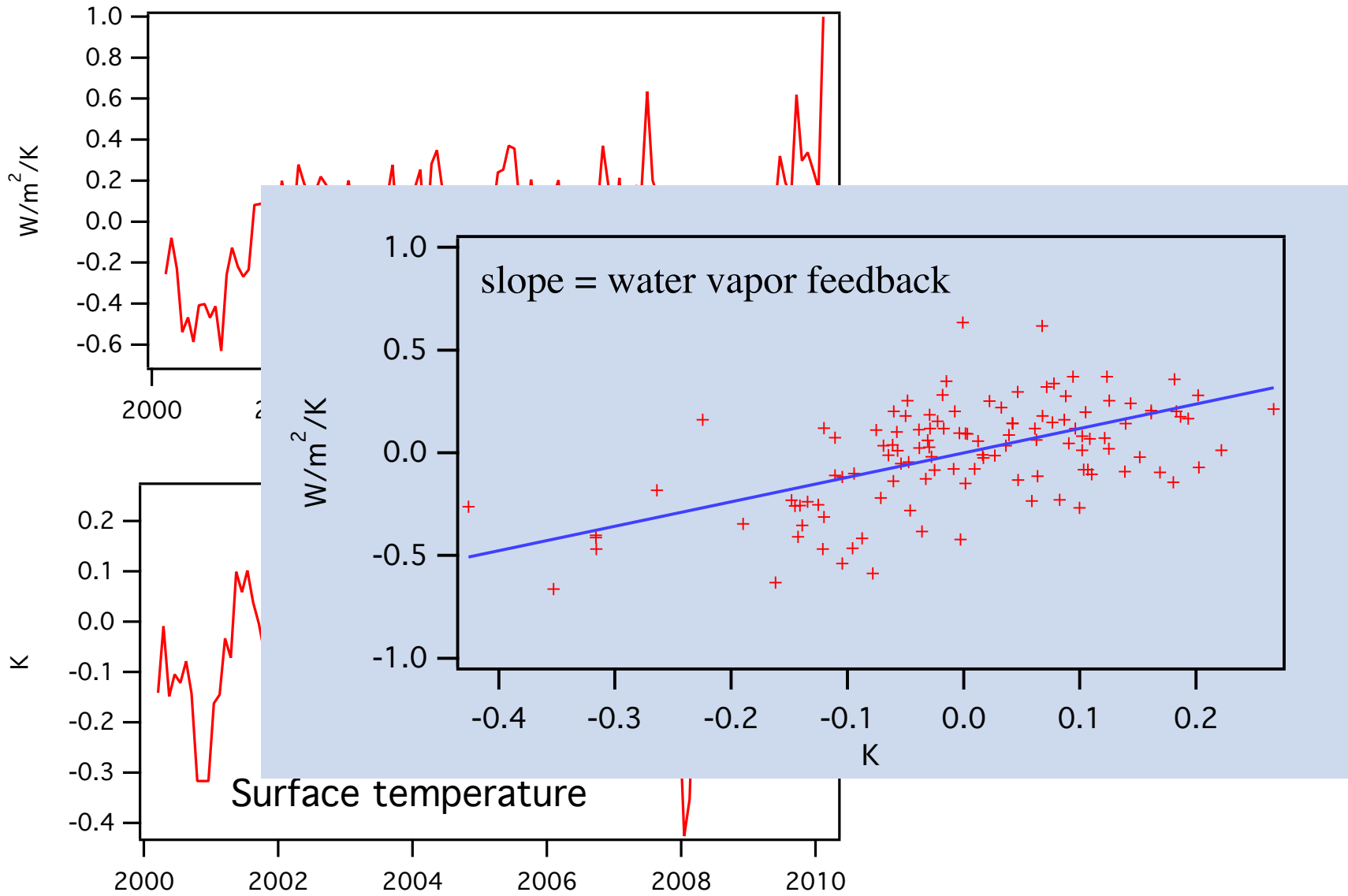
Regress energy trapped by e.g.,  $q$   
vs. surface temperature





Dessler et al., GRL, 2008

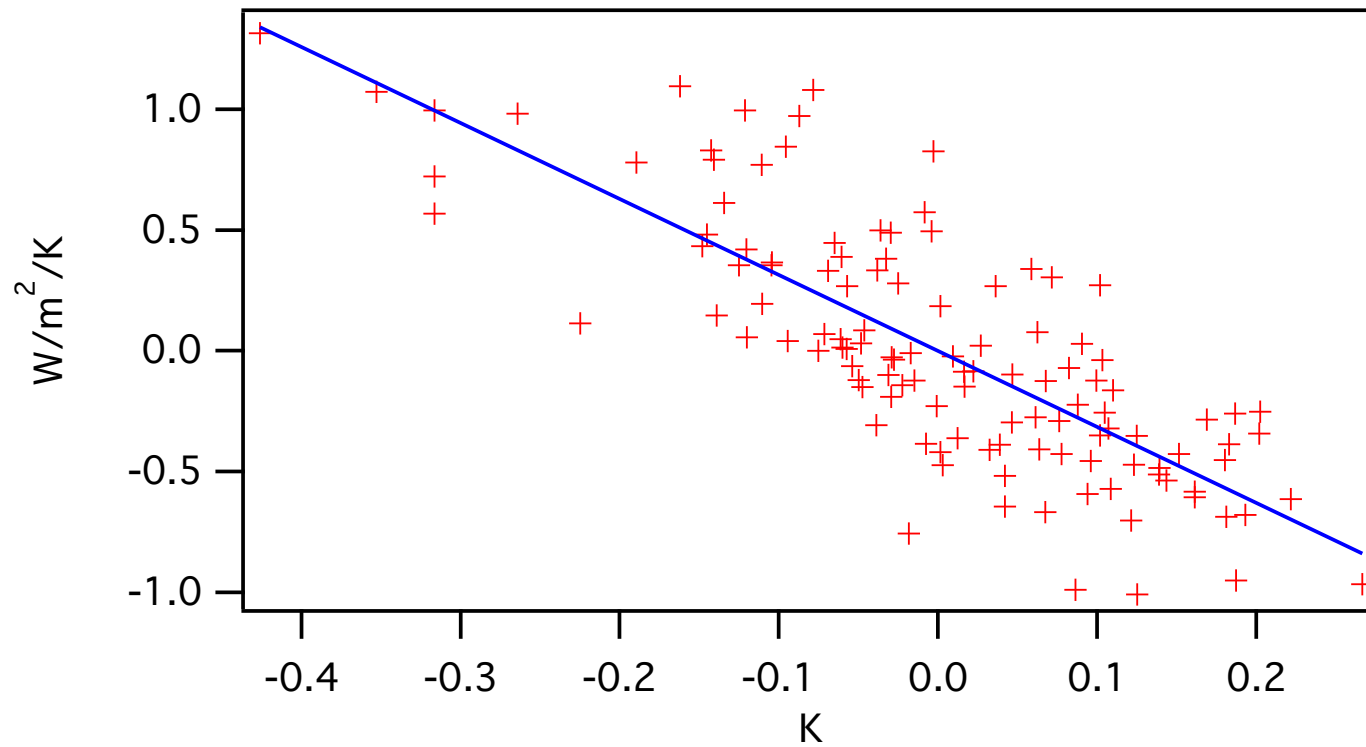
Dessler and Wong, J. Clim., 2009



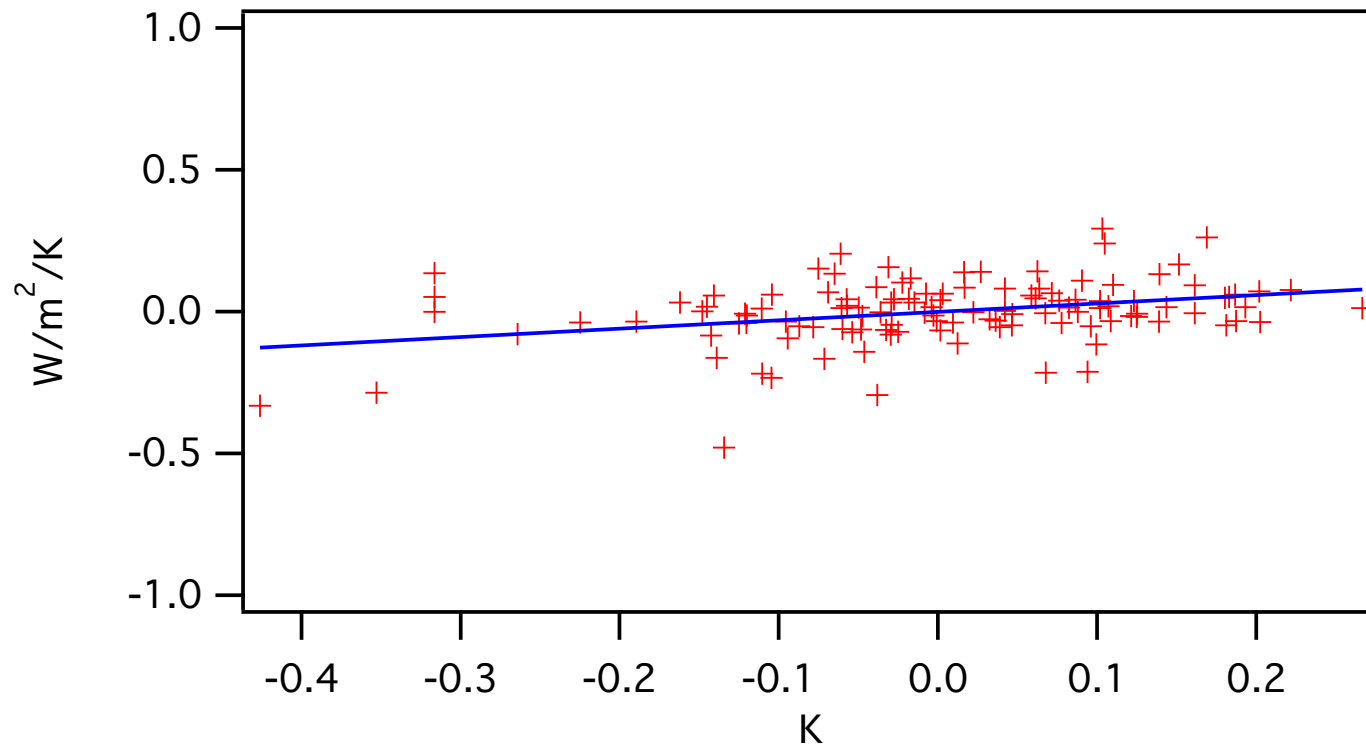
Dessler et al., GRL, 2008

Dessler and Wong, J. Clim., 2009

# Temperature feedback

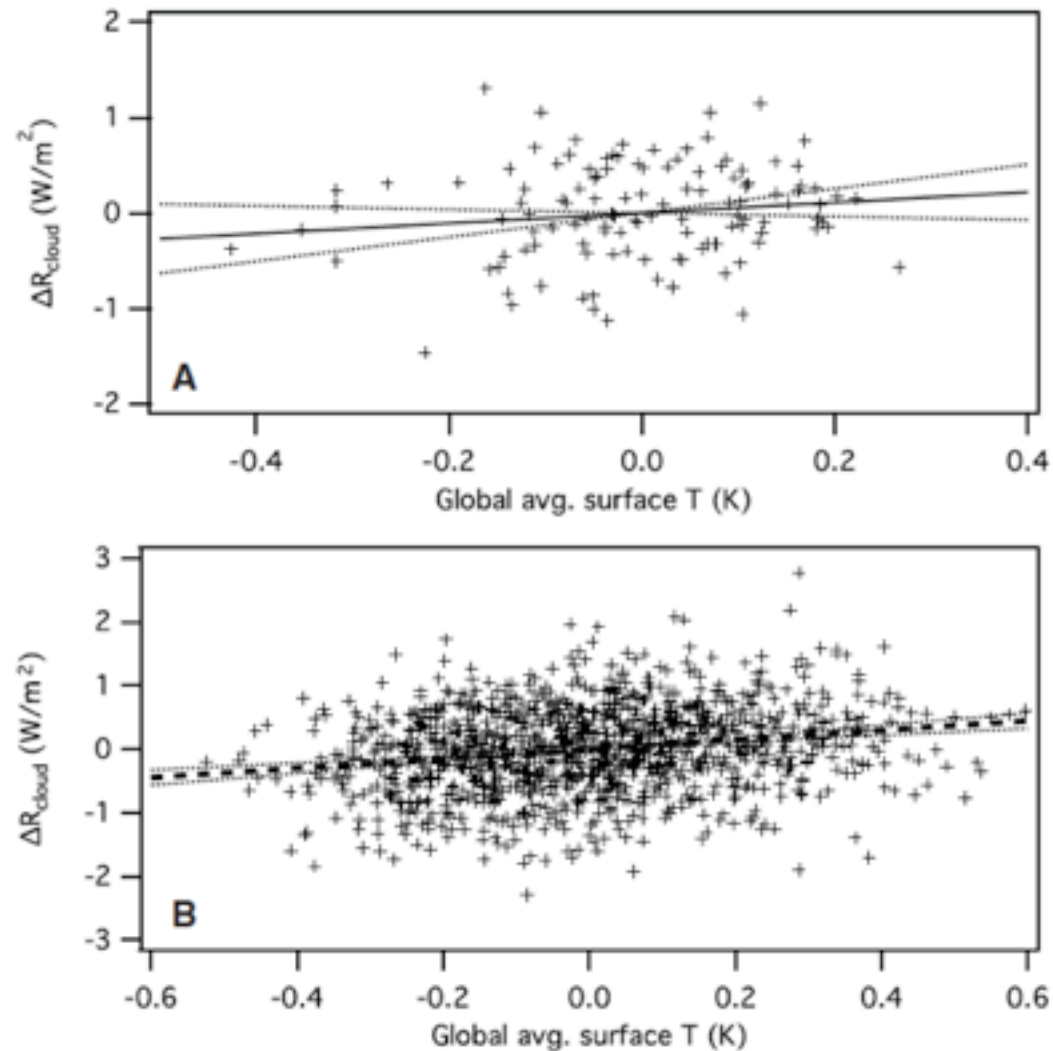


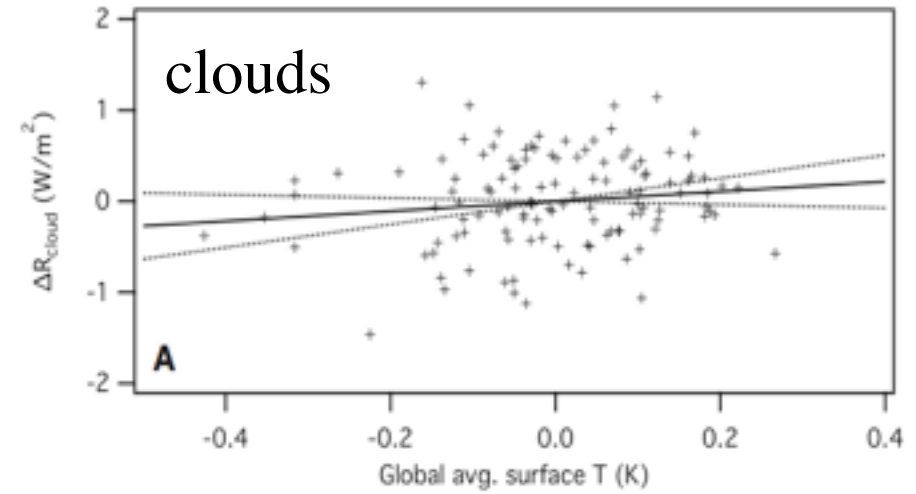
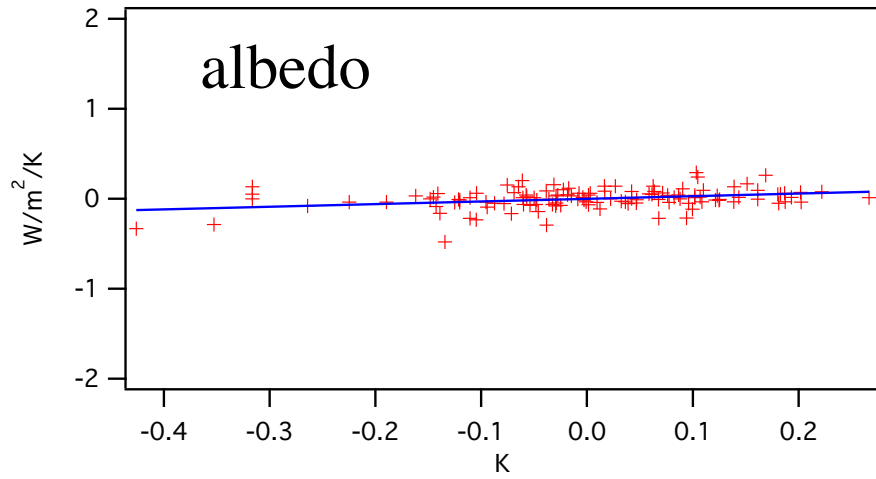
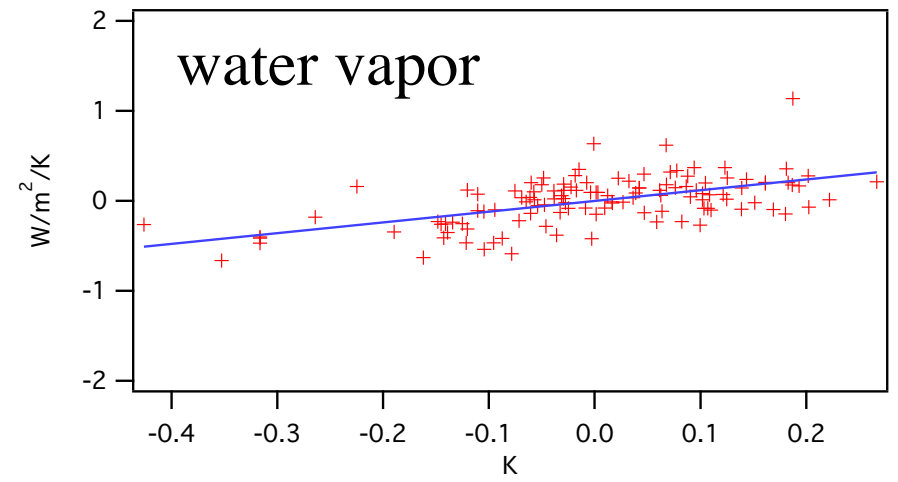
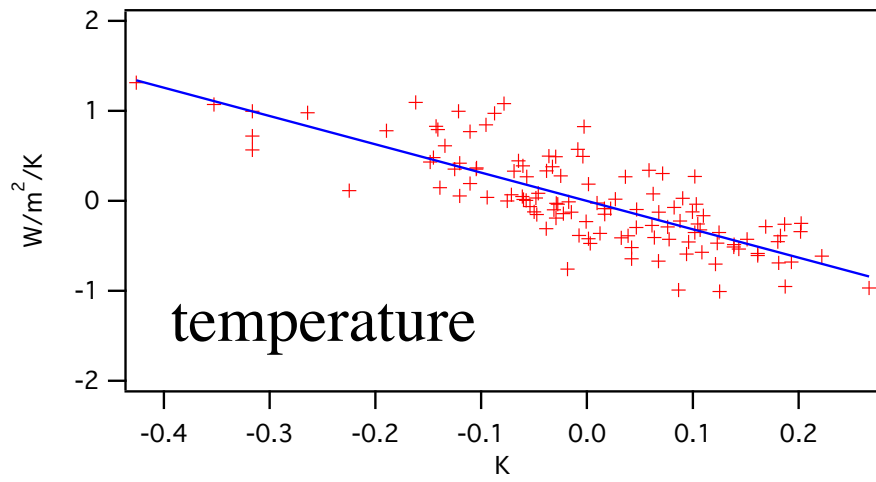
## albedo feedback

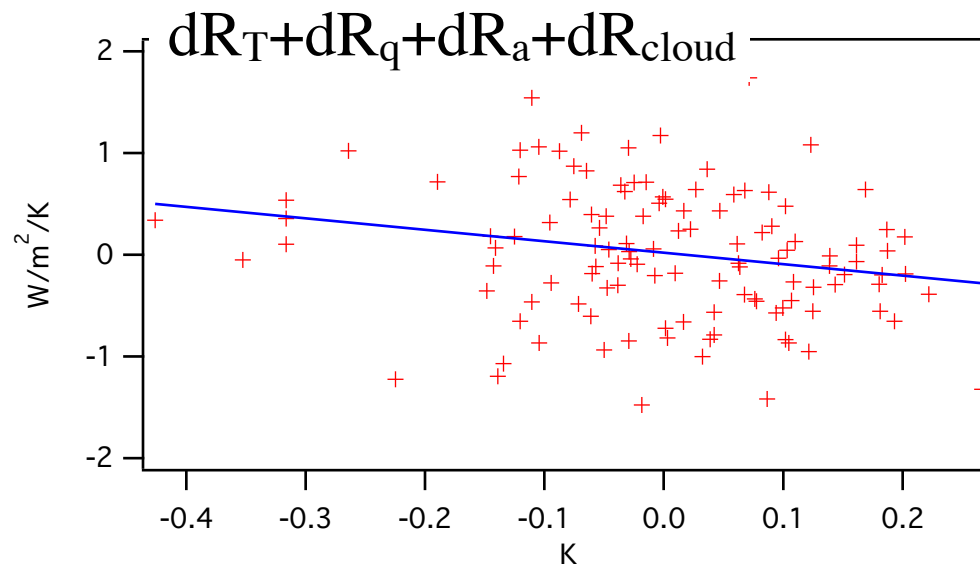
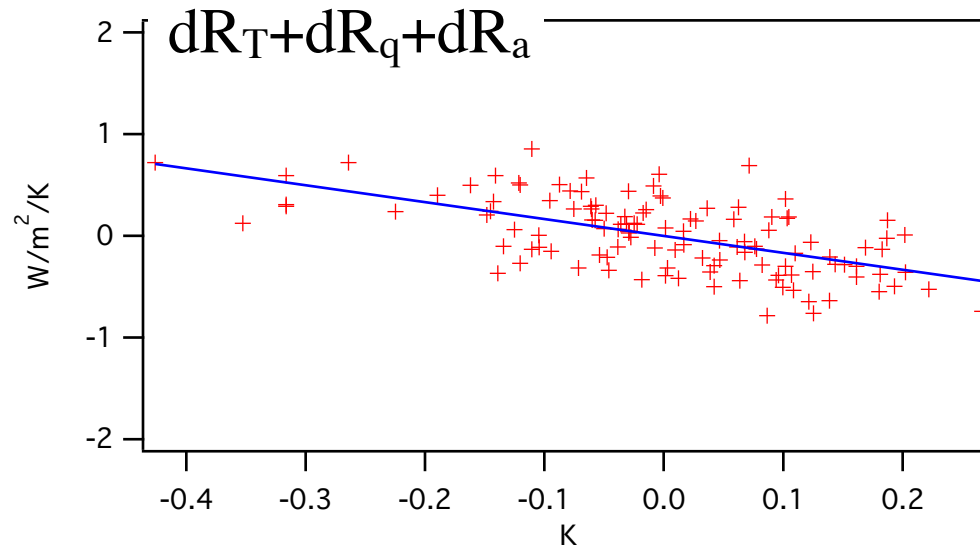


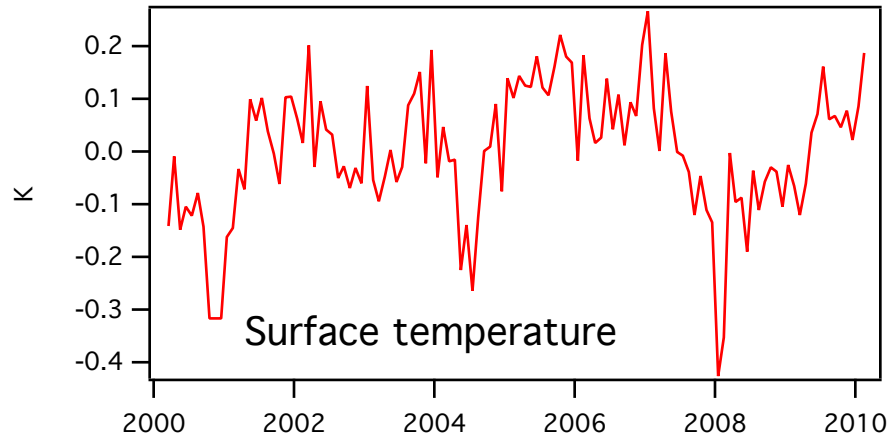
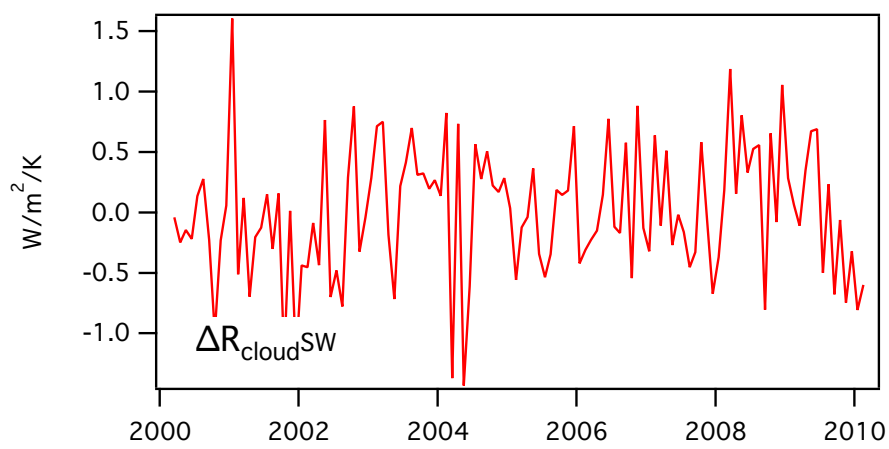
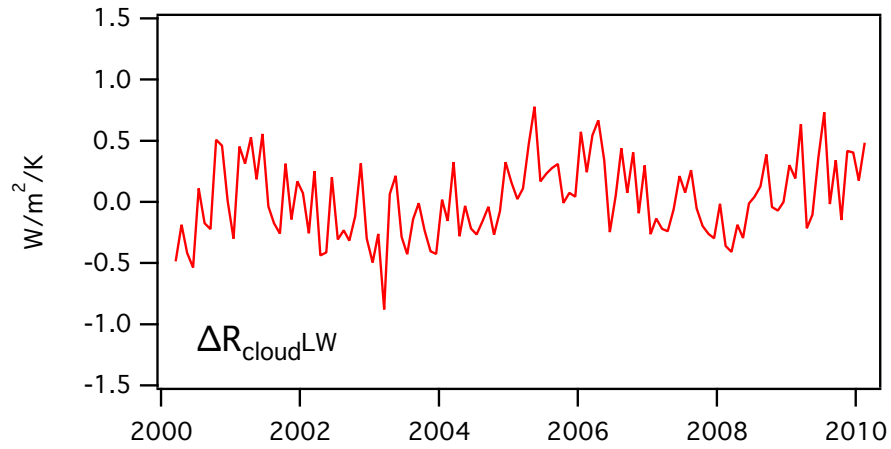
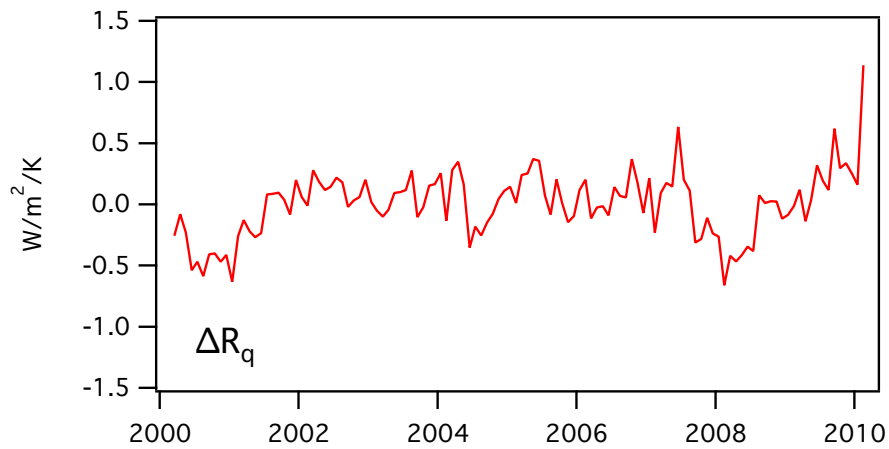
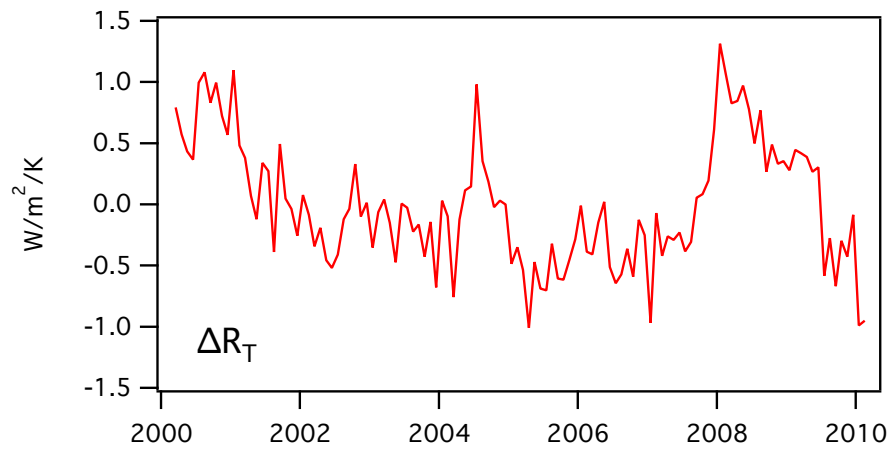


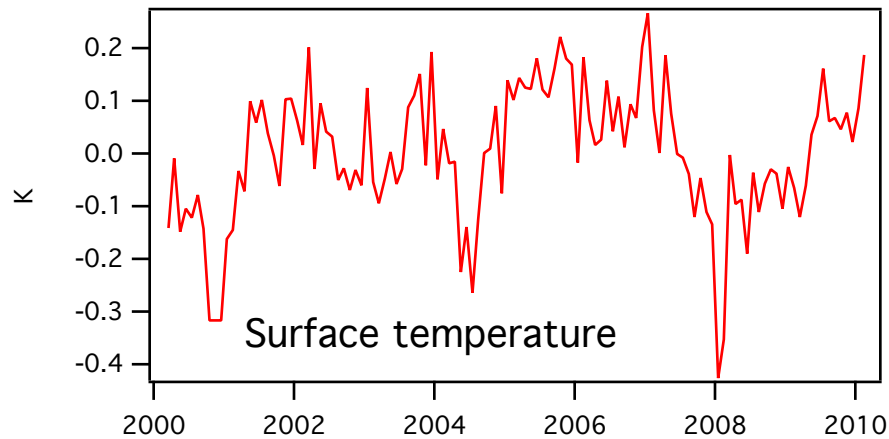
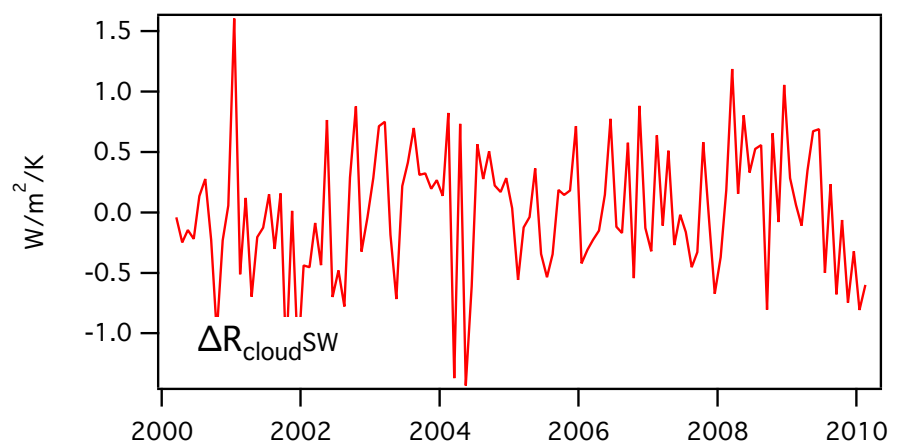
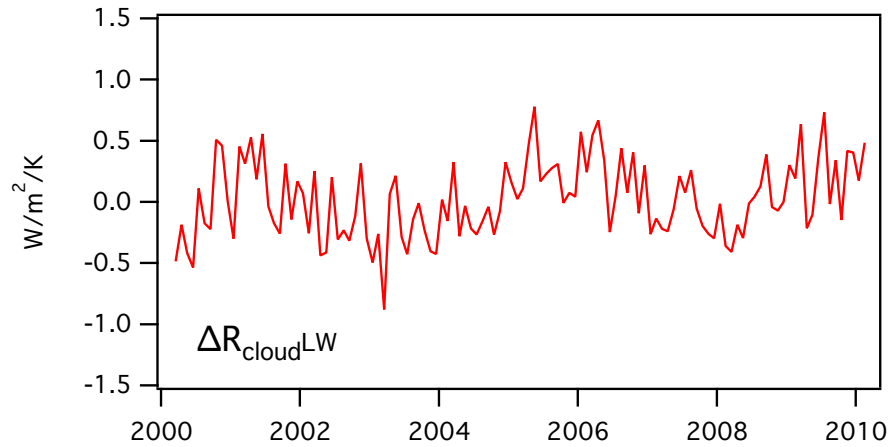
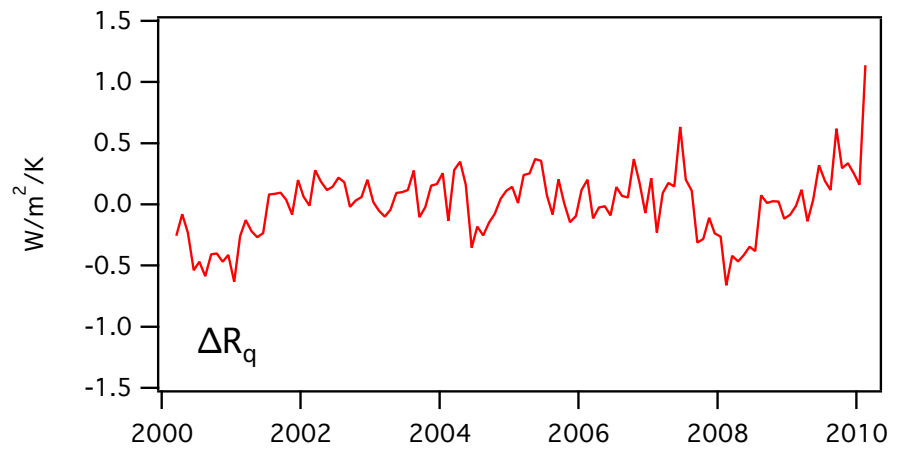
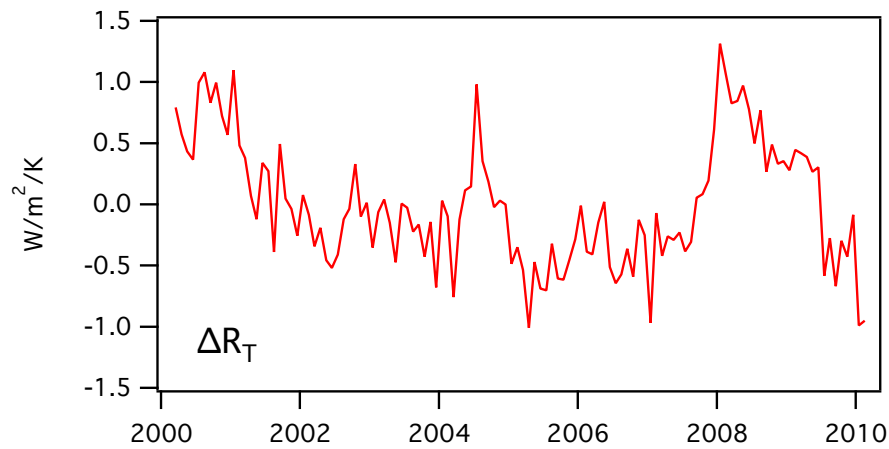
**Fig. 2.** (A) Scatter plot of monthly average values of  $\Delta R_{\text{cloud}}$  versus  $\Delta T_s$  using CERES and ECMWF interim data. (B) Scatter plot of monthly averages of the same quantities from 100 years of a control run of the ECHAM/MPI-OM model. In all plots, the solid line is a linear least-squares fit and the dotted lines are the  $2\sigma$  confidence interval of the fit.









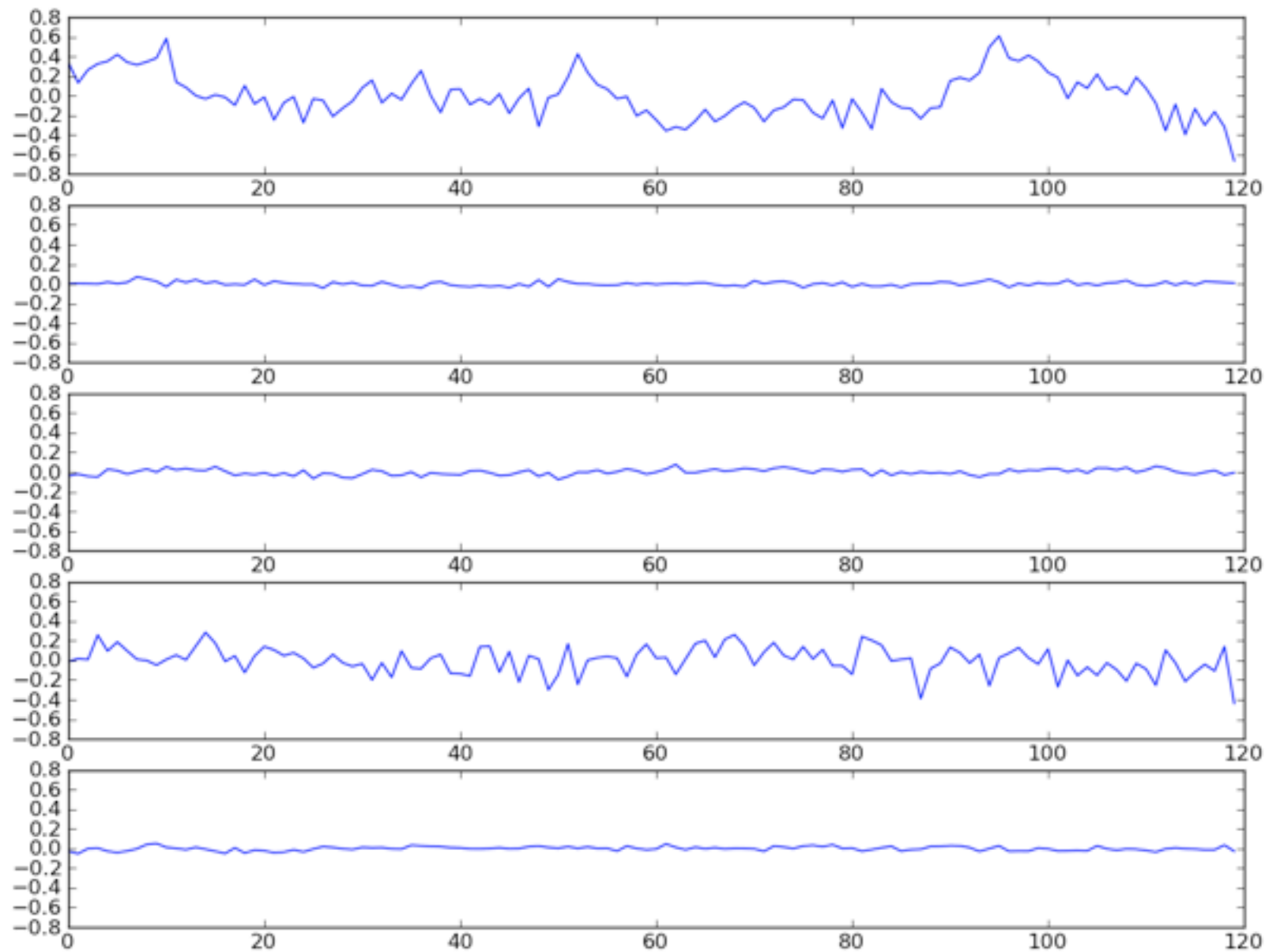


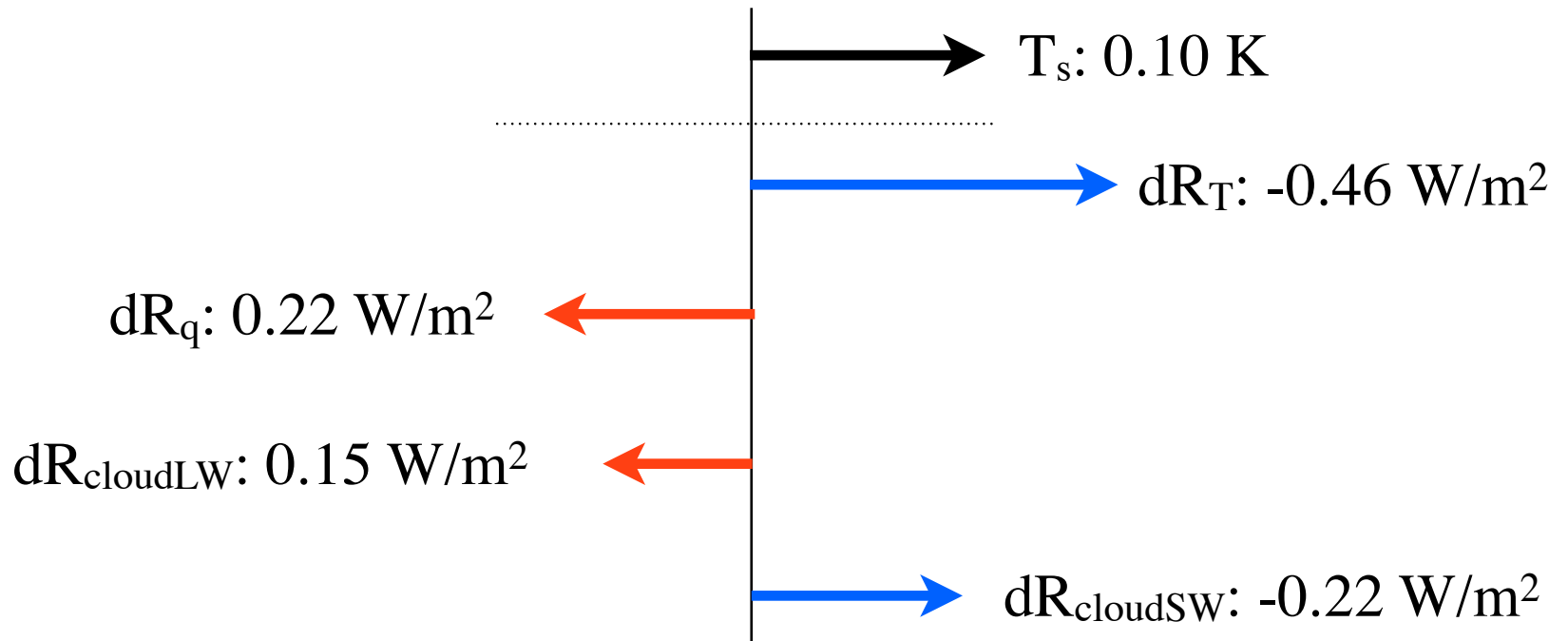
EOF analysis to find the  
modes of variability

# EOFs of water vapor time series

EOF

$dR_q$  (W/m<sup>2</sup>)



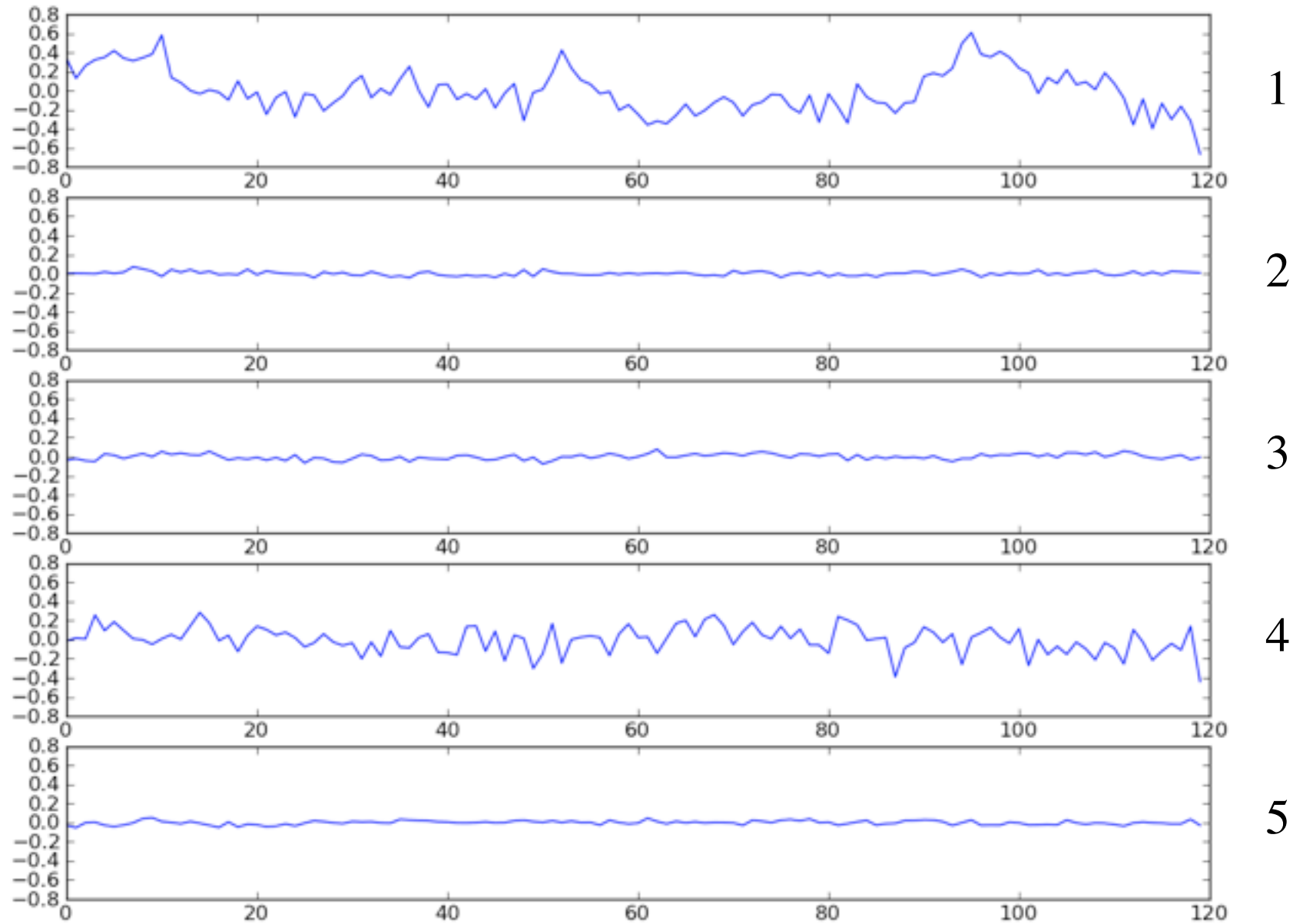


Explains 52% of variance

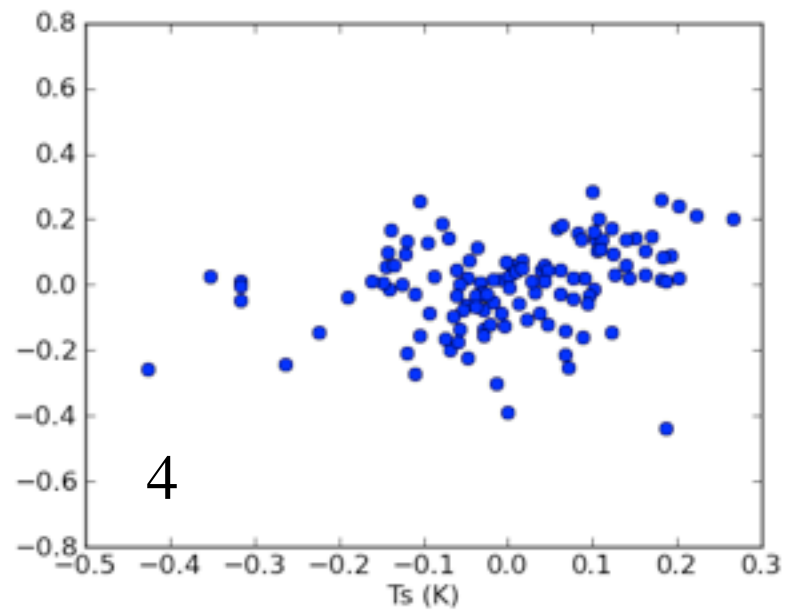
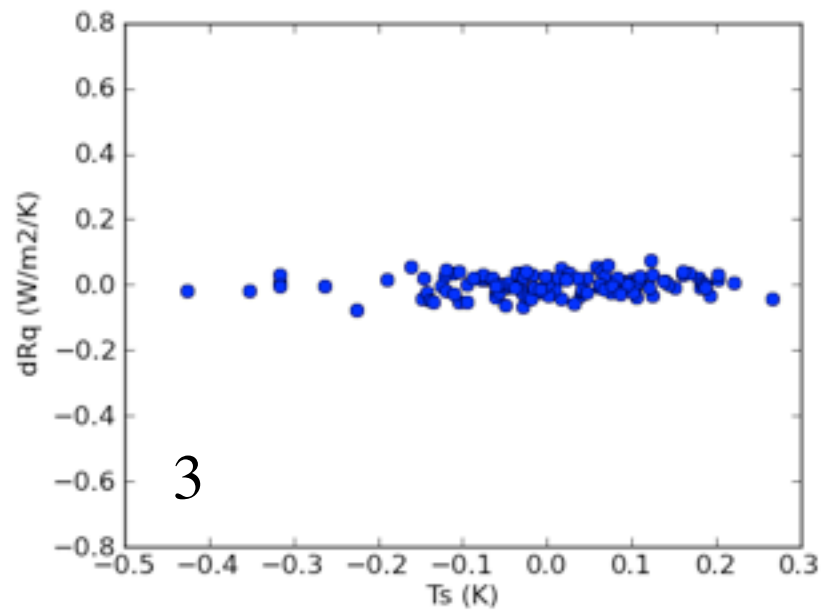
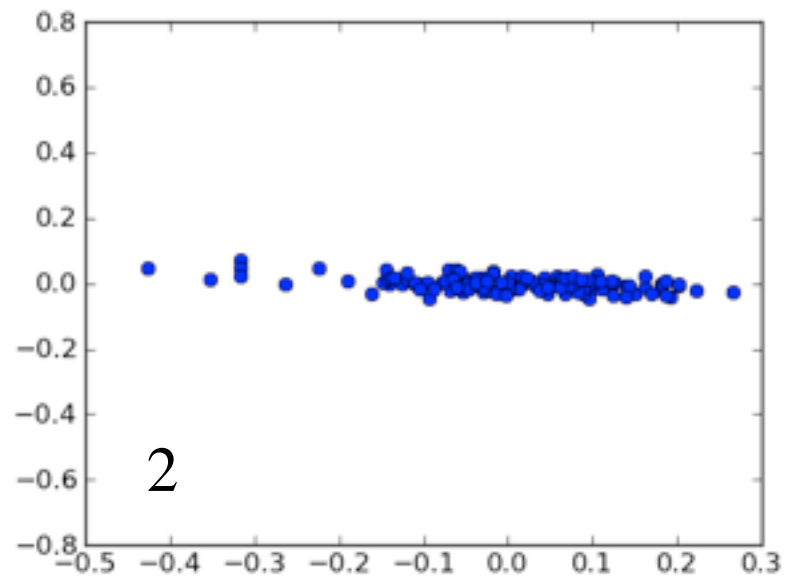
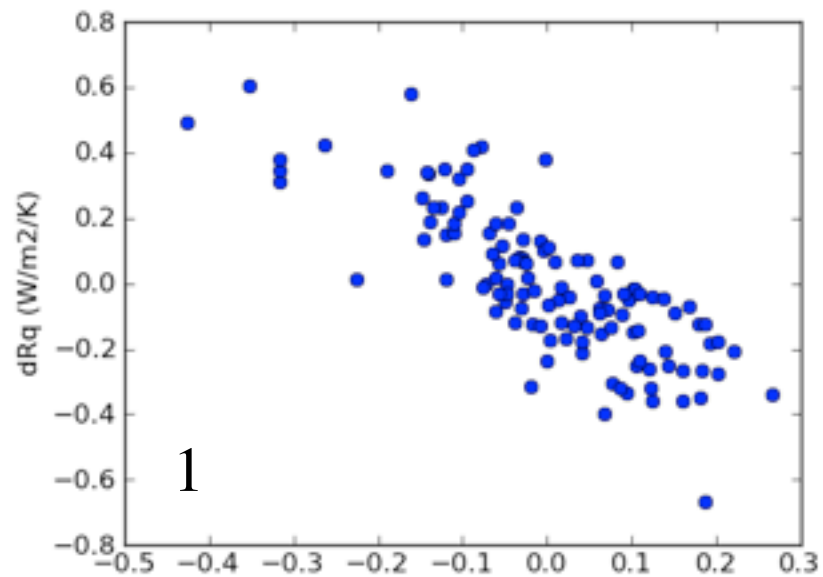
# EOFs of water vapor time series

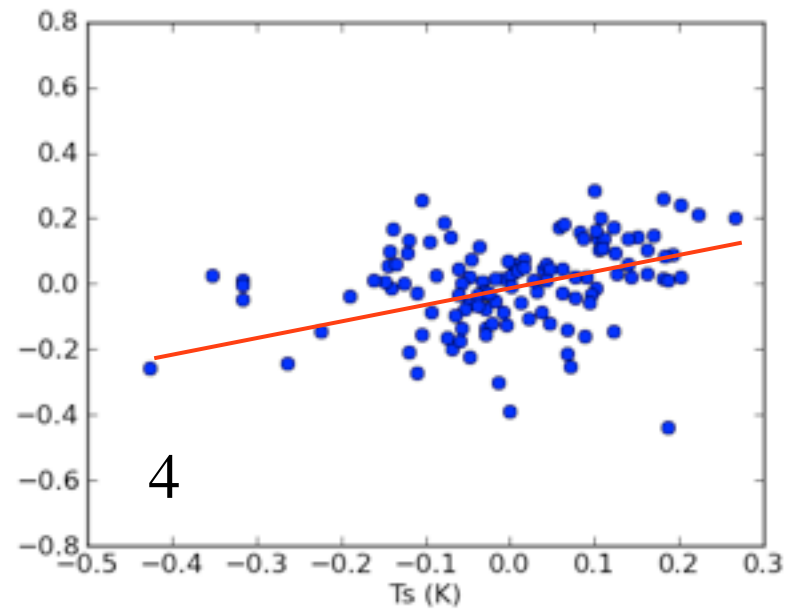
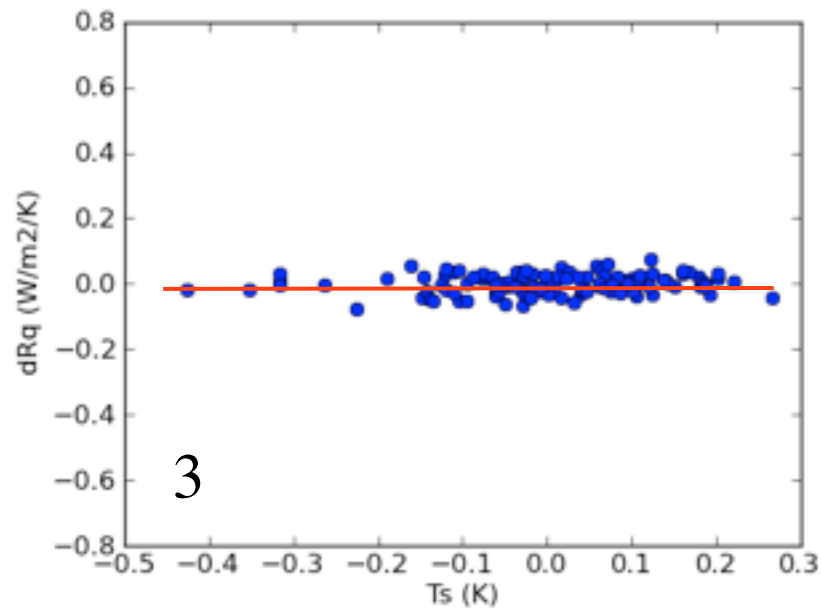
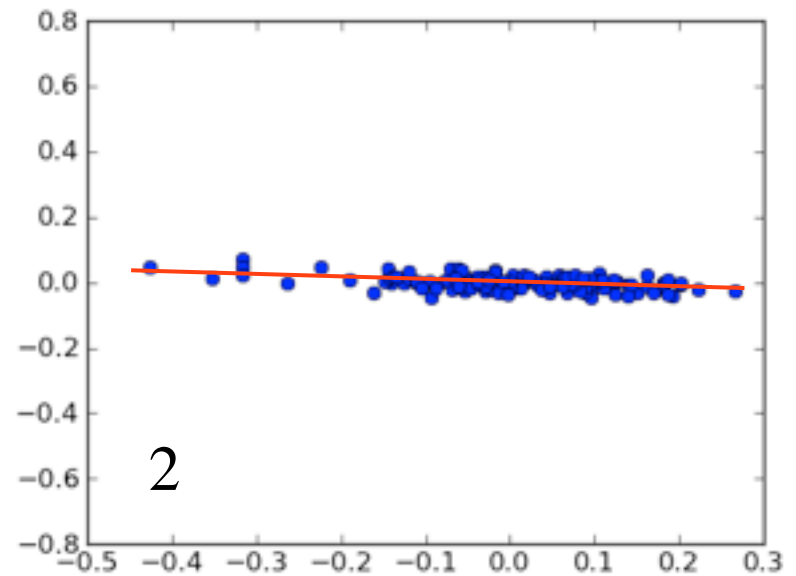
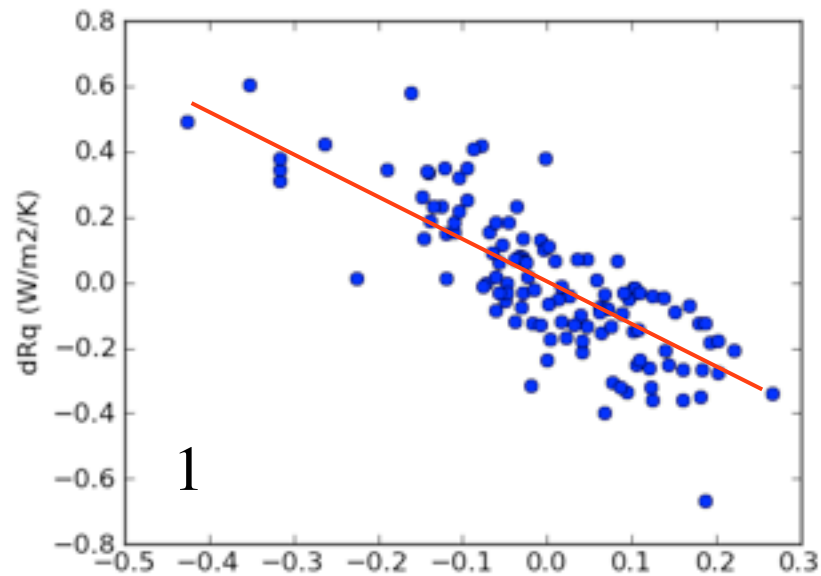
EOF

$dR_q$  (W/m<sup>2</sup>)









# Feedback

## $\text{W/m}^2/\text{K}$

52%

25%

12%

8%

3%

	EOF 1	EOF 2	EOF 3	EOF 4	EOF 5	total
$dR_T$	-2.91	-0.36	0.08	-0.17	0.24	-3.12
$dR_q$	1.42	0.08	-0.03	-0.32	0.04	1.18
$dR_{\text{cloudLW}}$	0.94	-0.71	0.22	-0.01	-0.02	0.42
$dR_{\text{cloudSW}}$	-1.36	1.47	0.29	-0.21	-0.08	0.11

tropical T

tropical +  
extratrop T

extratropical T

# Feedback

## $\text{W/m}^2/\text{K}$

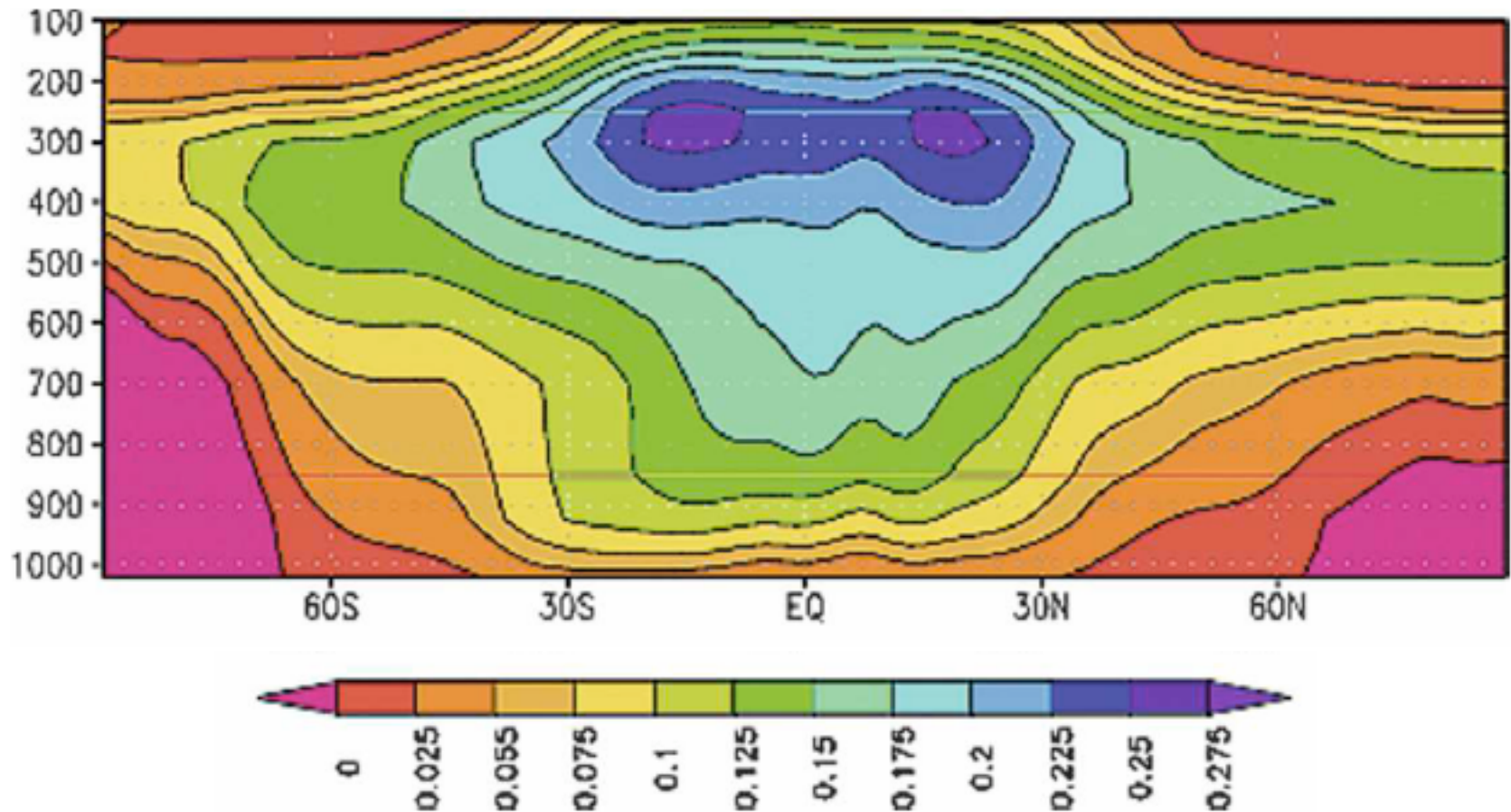
	EOF 1	EOF 2	EOF 3	EOF 4	EOF 5	total
$dR_T$	-2.91	-0.36	0.08	-0.17	0.24	-3.12
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↑  
tropical T

↑  
tropical +  
extratrop T

↙ ↑ ↘  
extratropical T

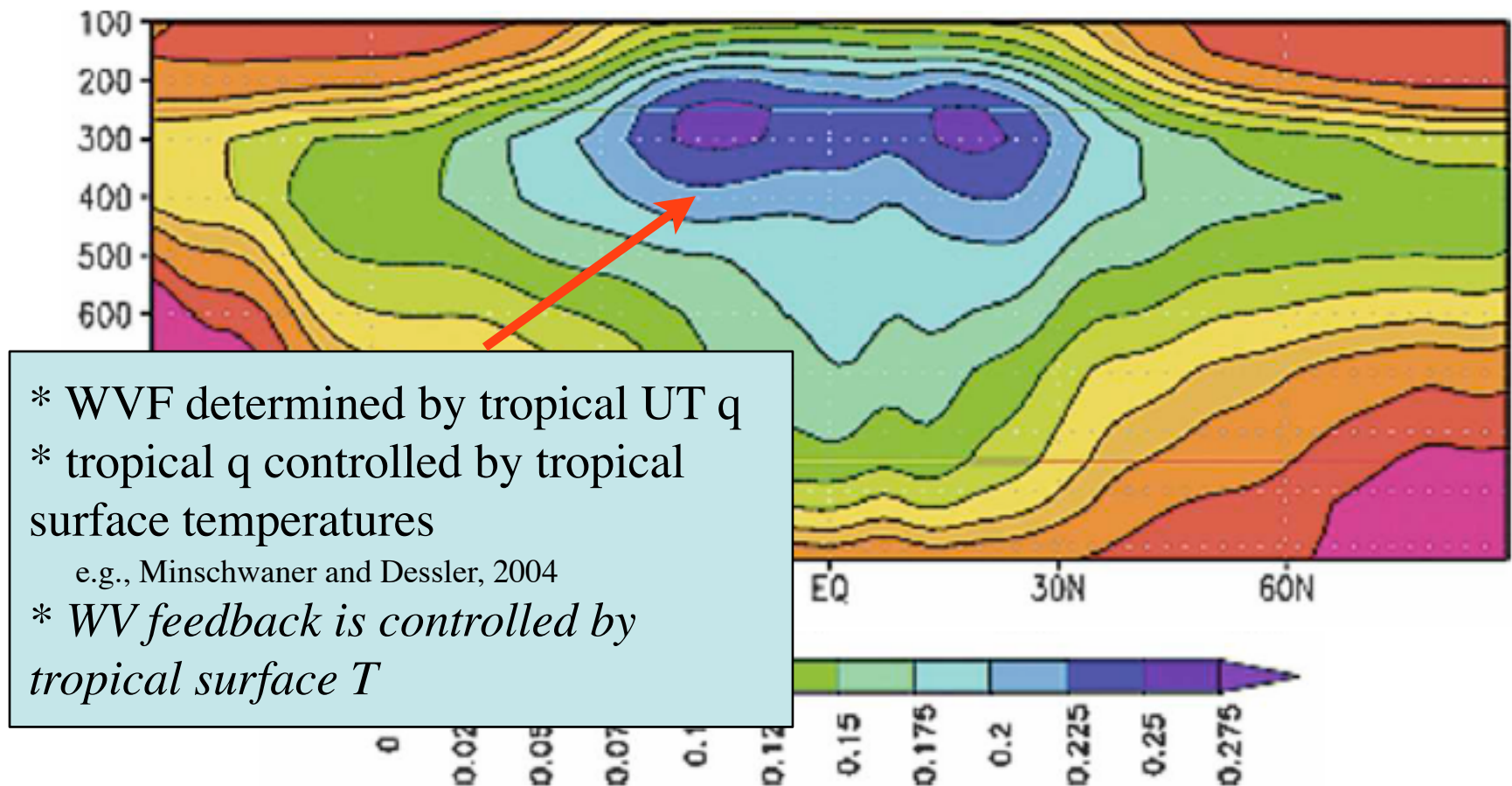
Water vapor feedback is primarily a “tropical” phenomenon



Change in R per unit change in  $q(x,y,z)$ :  $\Delta R/\Delta q(x,y,z)$

Fig. 2 of Soden et al., 2008

Water vapor feedback is primarily a “tropical” phenomenon



Change in  $R$  per unit change in  $q(x,y,z)$ :  $\Delta R / \Delta q(x,y,z)$

Fig. 2 of Soden et al., 2008

# Feedback

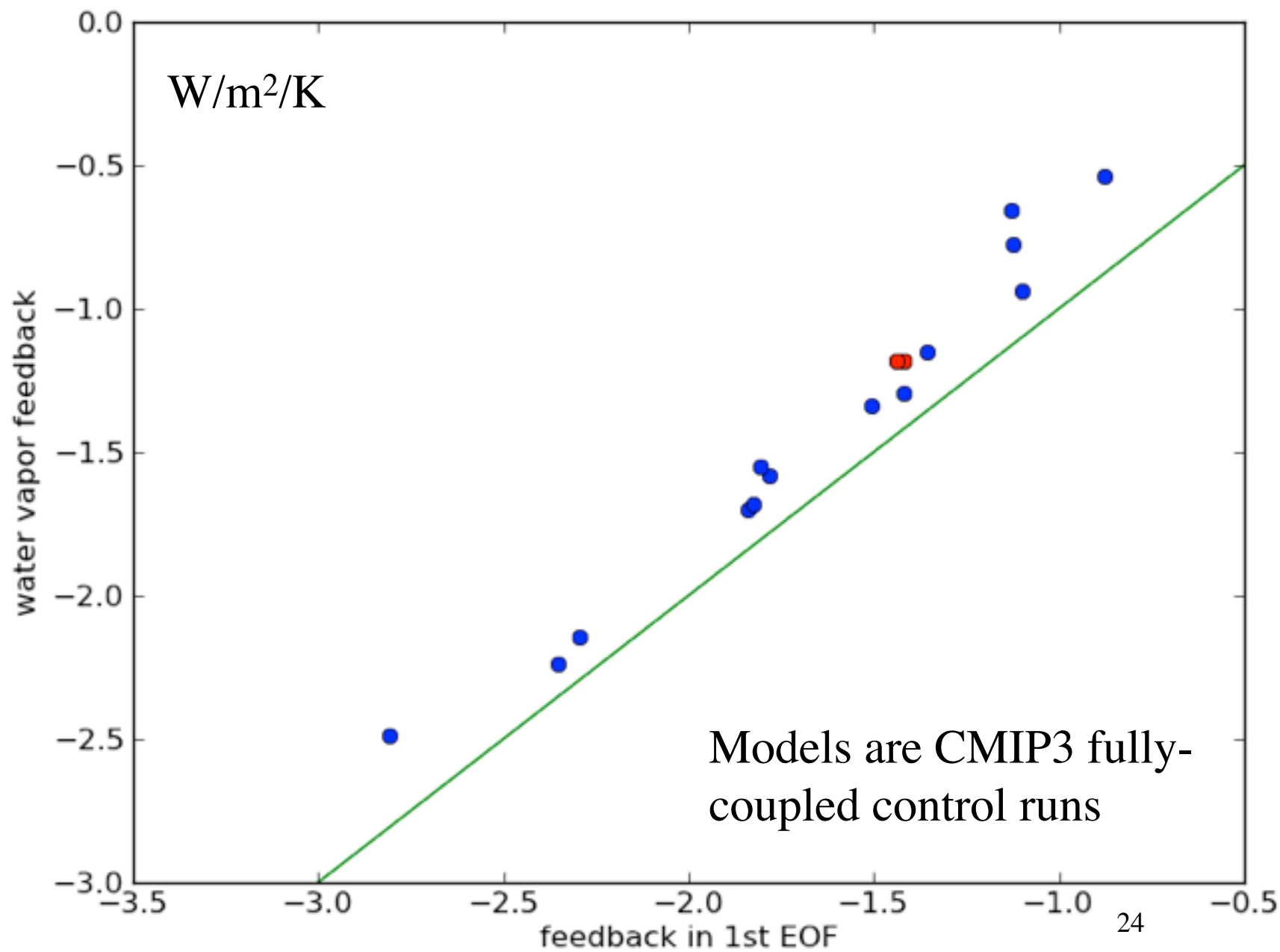
## $\text{W/m}^2/\text{K}$

	EOF 1	EOF 2	EOF 3	EOF 4	EOF 5	total
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↑  
tropical T

↑  
tropical +  
extratrop T

↙ ↑ ↘  
extratropical T



Models are CMIP3 fully-coupled control runs



# Feedback

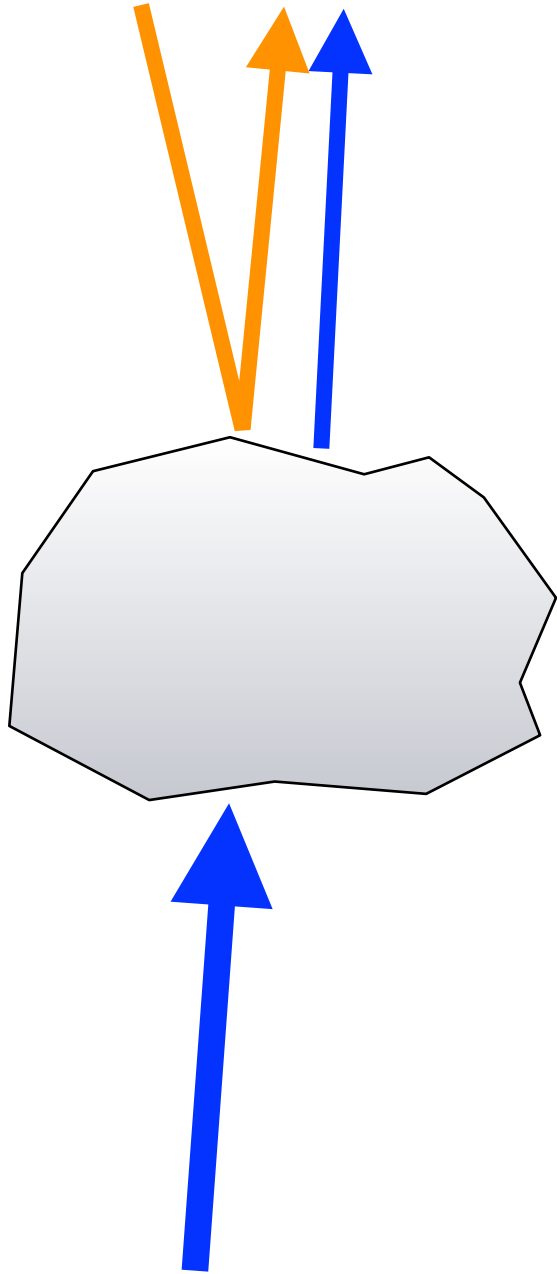
## $\text{W/m}^2/\text{K}$

	EOF 1	EOF 2	EOF 3	EOF 4	EOF 5	total
$dR_T$	-2.91	-0.36	0.08	-0.17	0.24	-3.12
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↑  
tropical T

↑  
tropical +  
extratrop T

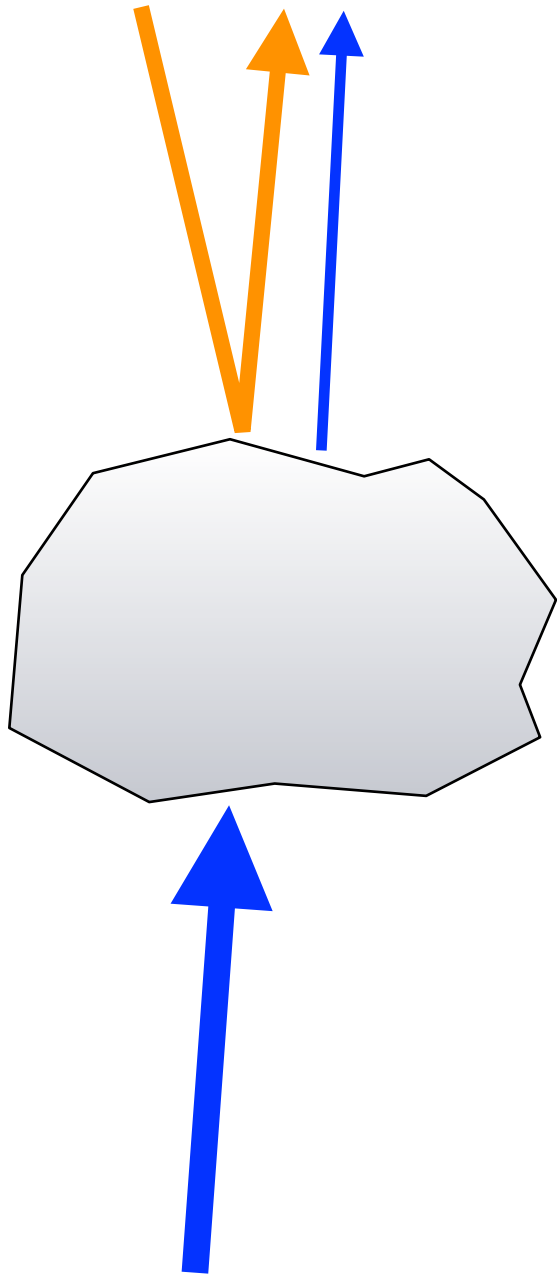
↙ ↑ ↘  
extratropical T



## Effect of clouds on top-of-atmosphere (TOA) flux

- 1) reduce incoming solar: cool
- 2) reduce outgoing IR: warm

net effect is the difference between these effects

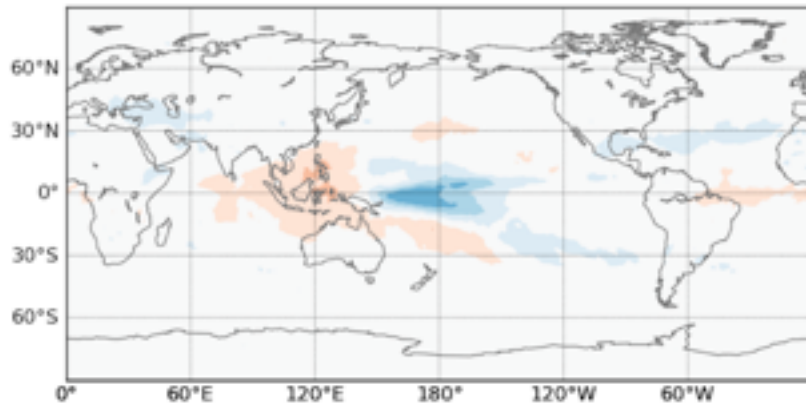


in today's atmosphere, clouds reduce net energy in to the Earth by  $20 \text{ W/m}^2$  (also known as cloud radiative forcing)

how will this change in a future climate?

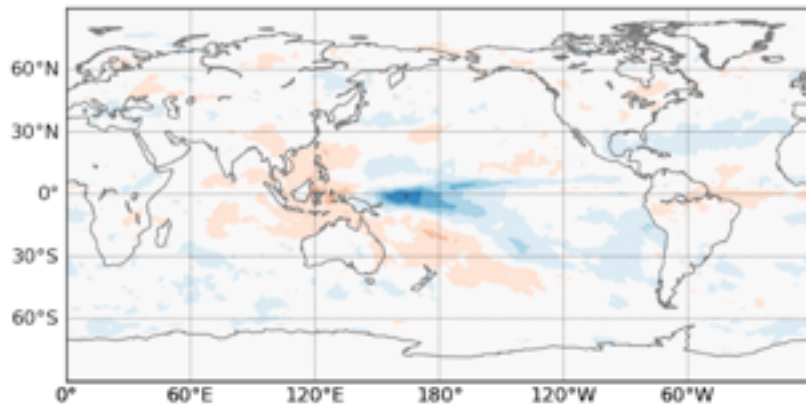
if changing clouds further reduce TOA downward net flux, this is a negative feedback

if changing clouds increase TOA downward net flux, this is a positive feedback



LW EOF 1

*Covariance of PC1 vs.  
time series at each grid  
point of LW and SW energy  
trapped by clouds*

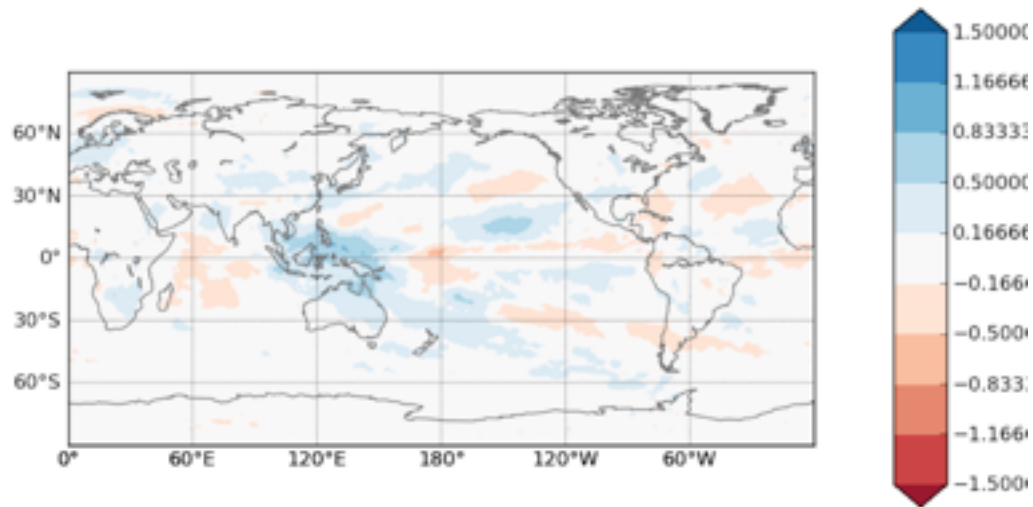


SW EOF 1

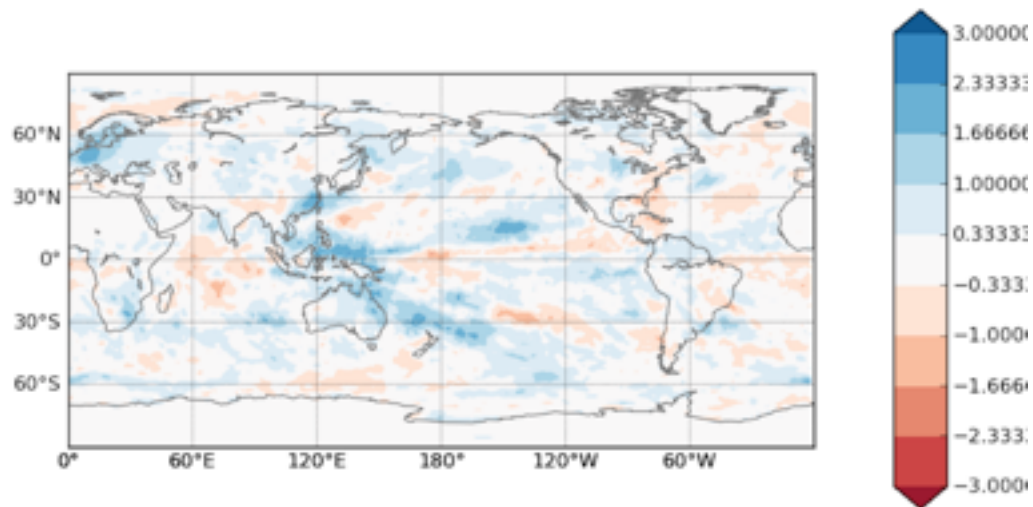
# Conclusions

- Clouds that make it difficult to accurately determine how TOA flux anomaly varies with surface temperature
  - they correlate poorly with surface T
  - next steps: use EOF analysis to gain insight into the factors that regulate clouds
  - goal is to improve estimate of clouds vs. T
- Water vapor and temperature are well behaved

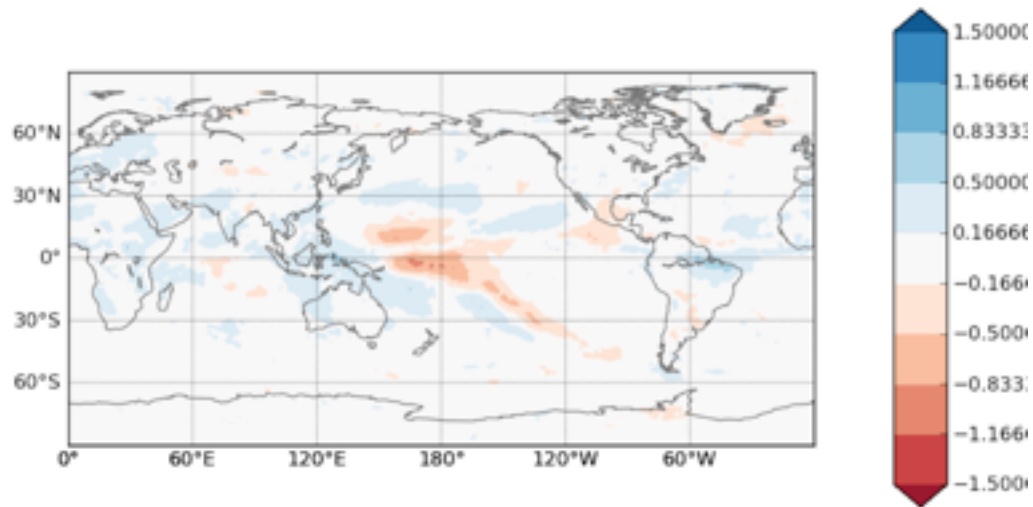




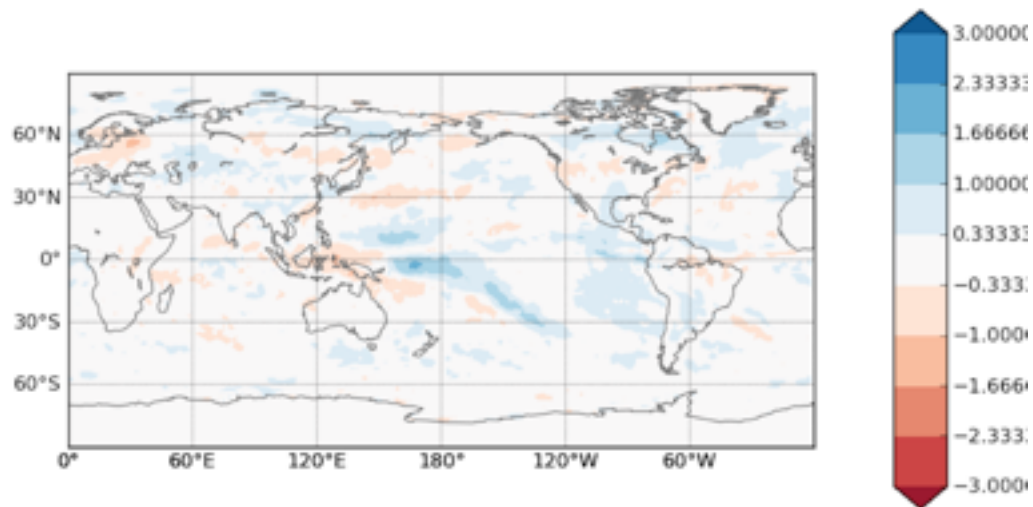
LW EOF 2



SW EOF 2



LW EOF 3



SW EOF 3

# Feedback

## $\text{W/m}^2/\text{K}$

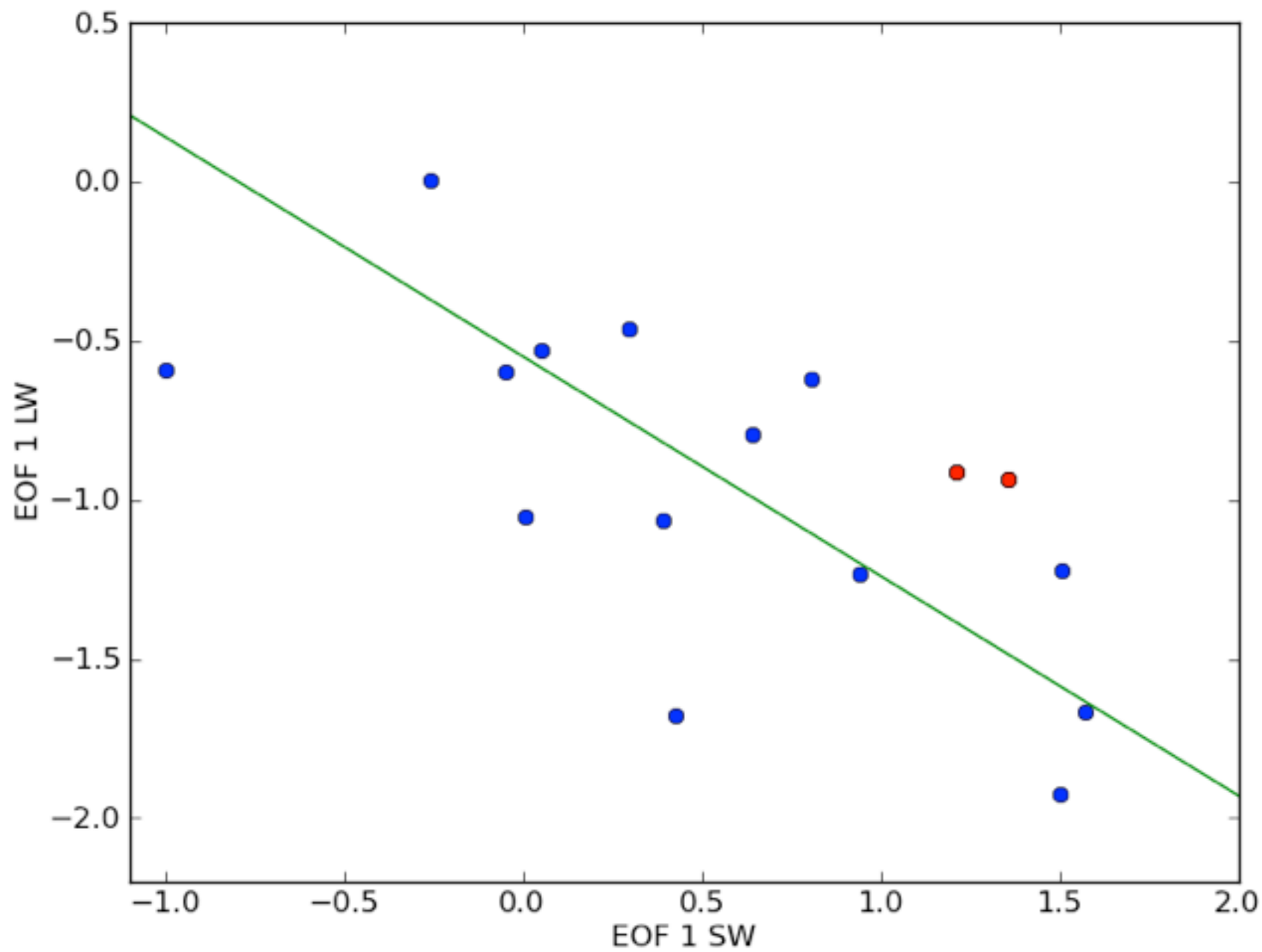
	EOF 1	EOF 2	EOF 3	EOF 4	EOF 5	total
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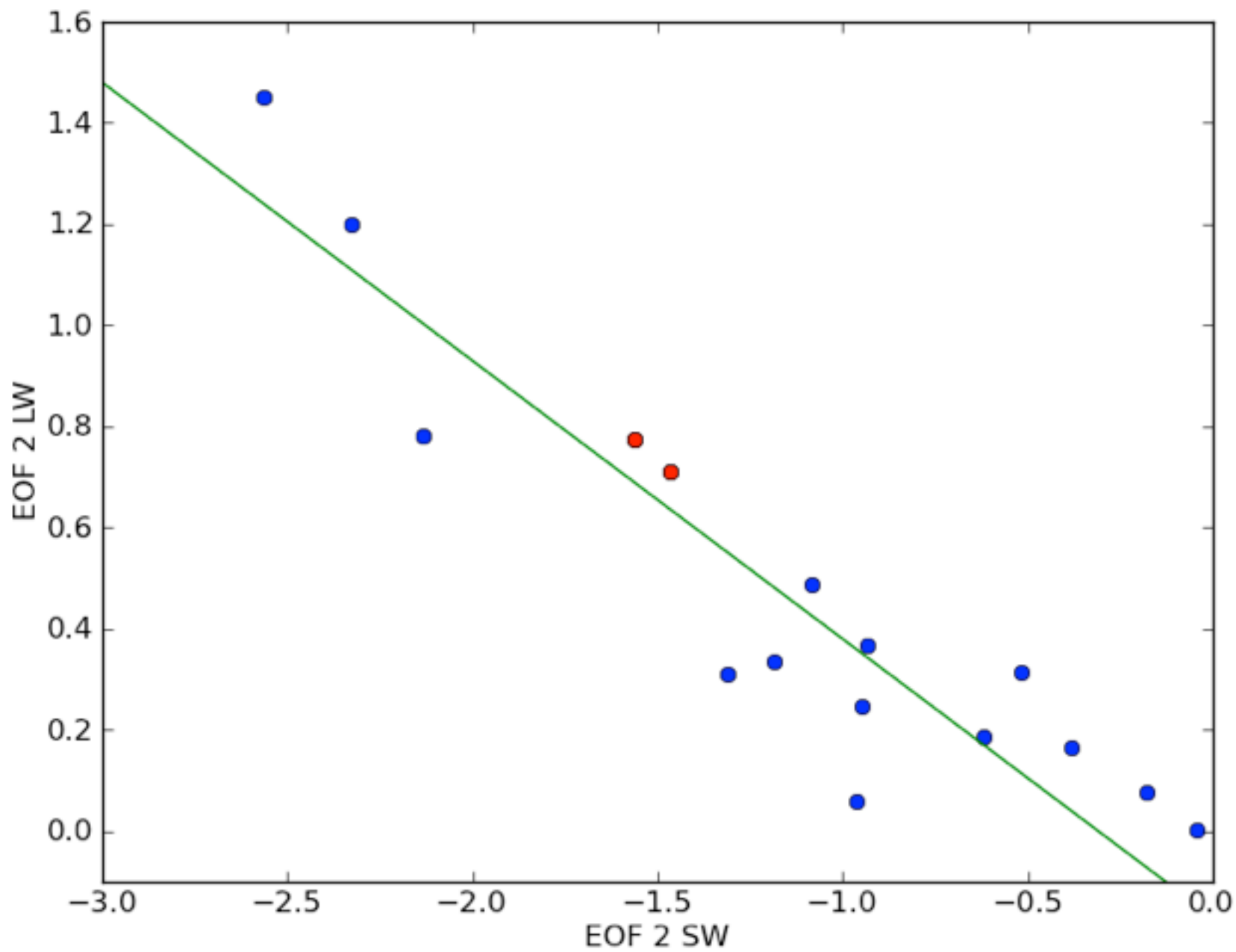
↑  
tropical T

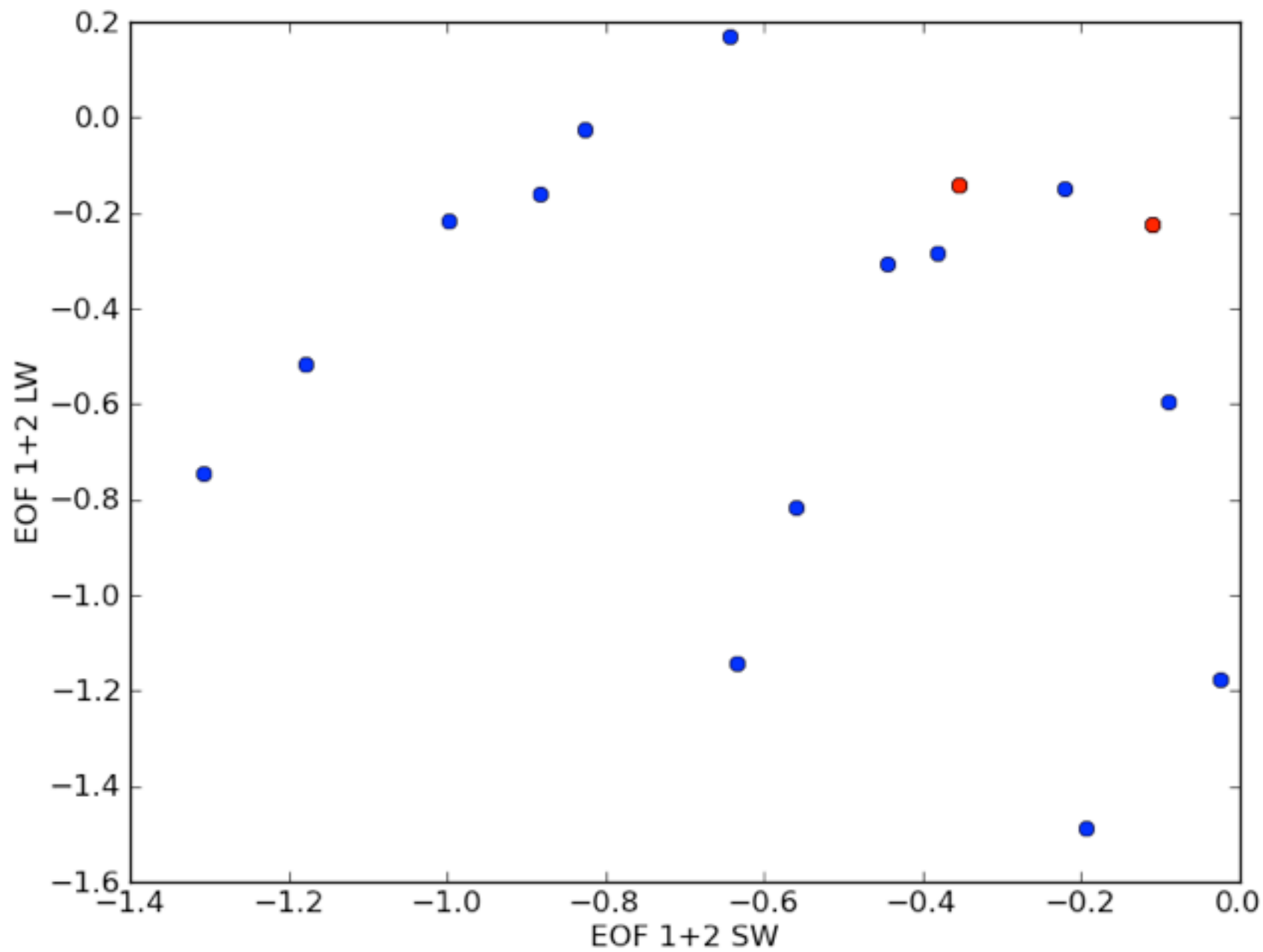
↑  
tropical +  
extratrop T

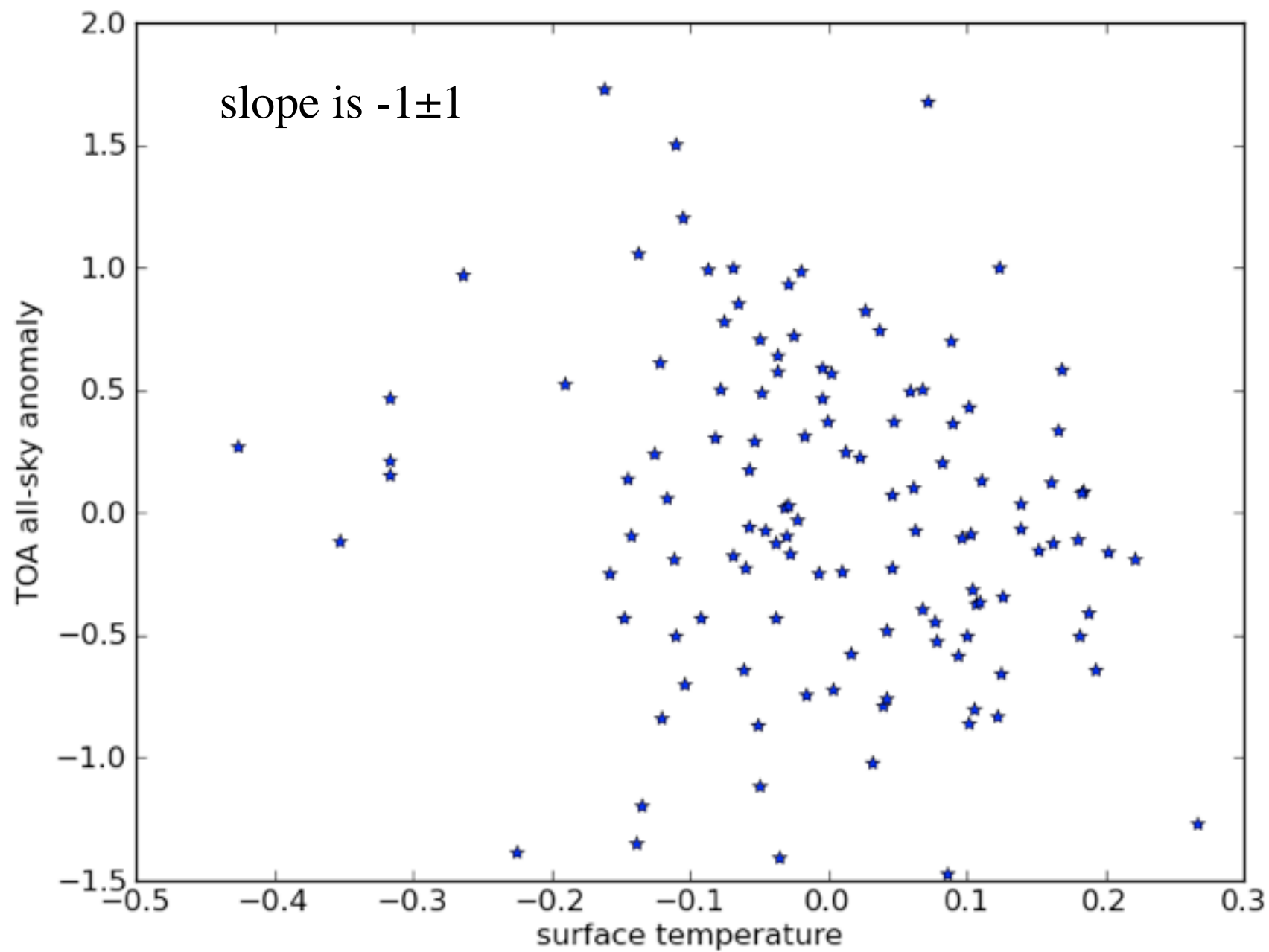
↙ ↑ ↘  
extratropical T











to determine  $\Delta R_{\text{cloud}}$

- start with cloud radiative forcing ( $\Delta \text{CRF}$ ); change in TOA flux if clouds are removed
- $\Delta \text{CRF} = (\Delta R_{\text{clear-sky}} - \Delta R_{\text{all-sky}})$
- $\Delta \text{CRF}$  can also be affected by changes in  $T$ ,  $q$ , albedo, radiative forcing
- Soden et al. [2008] adjustment to get  $\Delta R_{\text{cloud}}$  from  $\Delta \text{CRF}$ ; see also Shell et al. [2008]

$$\Delta R_{\text{cloud}} = \Delta \text{CRF} + (K^0_T - K_T)dT + (K^0_W - K_W)dW \\ + (K^0_a - K_a)da + (G^0 - G).$$

$$\Delta R_{cloud} = \left( \Delta R_{clear-sky} - \Delta R_{all-sky} \right) + (K^0_T - K_T)dT + (K^0_W - K_W)dW \\ + (K^0_a - K_a)da + (G^0 - G).$$

cloud radiative forcing



$$\Delta R_{cloud} = \left( \Delta R_{clear-sky} - \Delta R_{all-sky} \right) + (K_T^0 - K_T)dT + (K_W^0 - K_W)dW \\ + (K_a^0 - K_a)da + (G^0 - G).$$

cloud radiative forcing

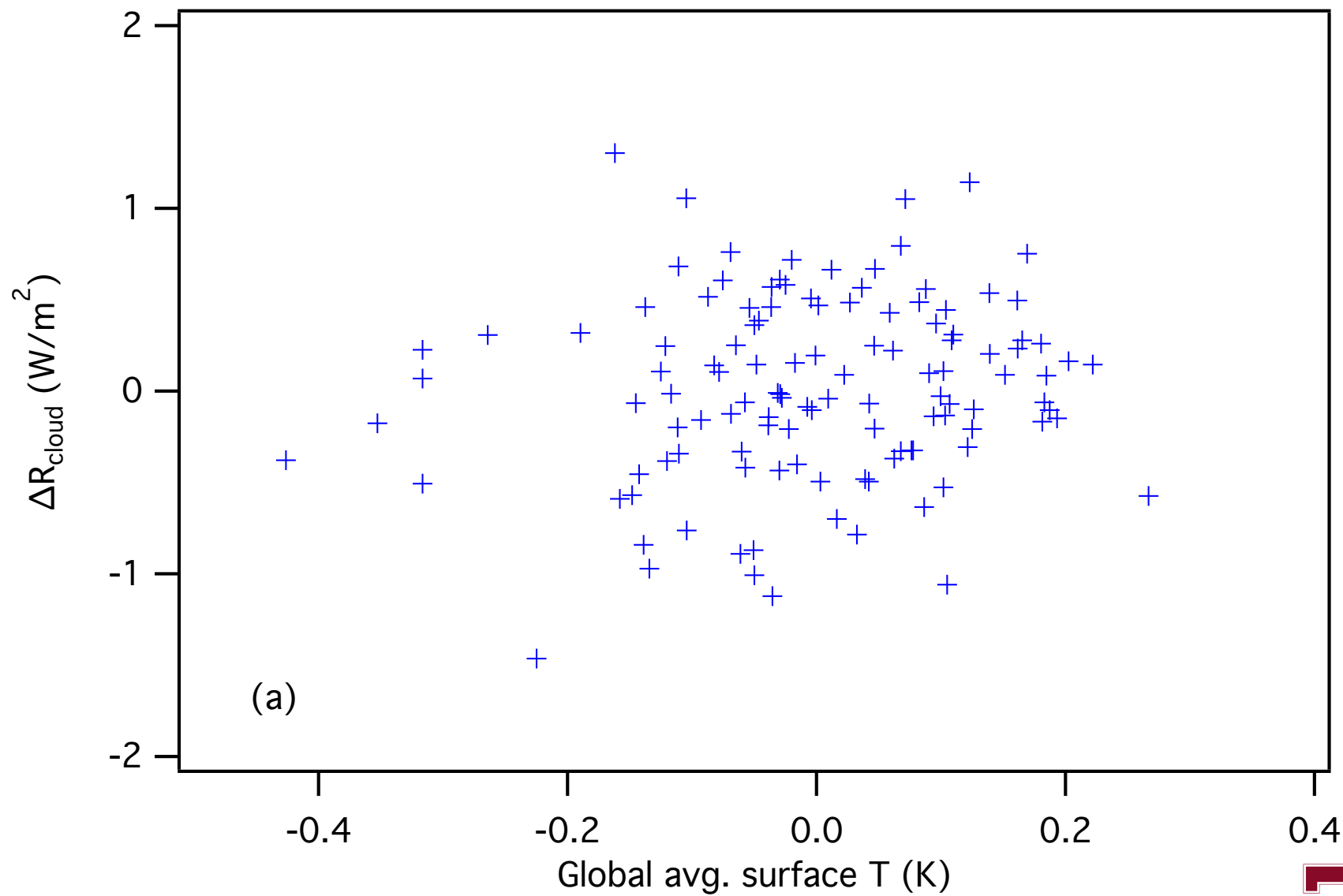


$$\Delta R_{cloud} = \left( \Delta R_{clear-sky} - \Delta R_{all-sky} \right) + (K_T^0 - K_T)dT + (K_W^0 - K_W)dW \\ + (K_a^0 - K_a)da + (G^0 - G).$$

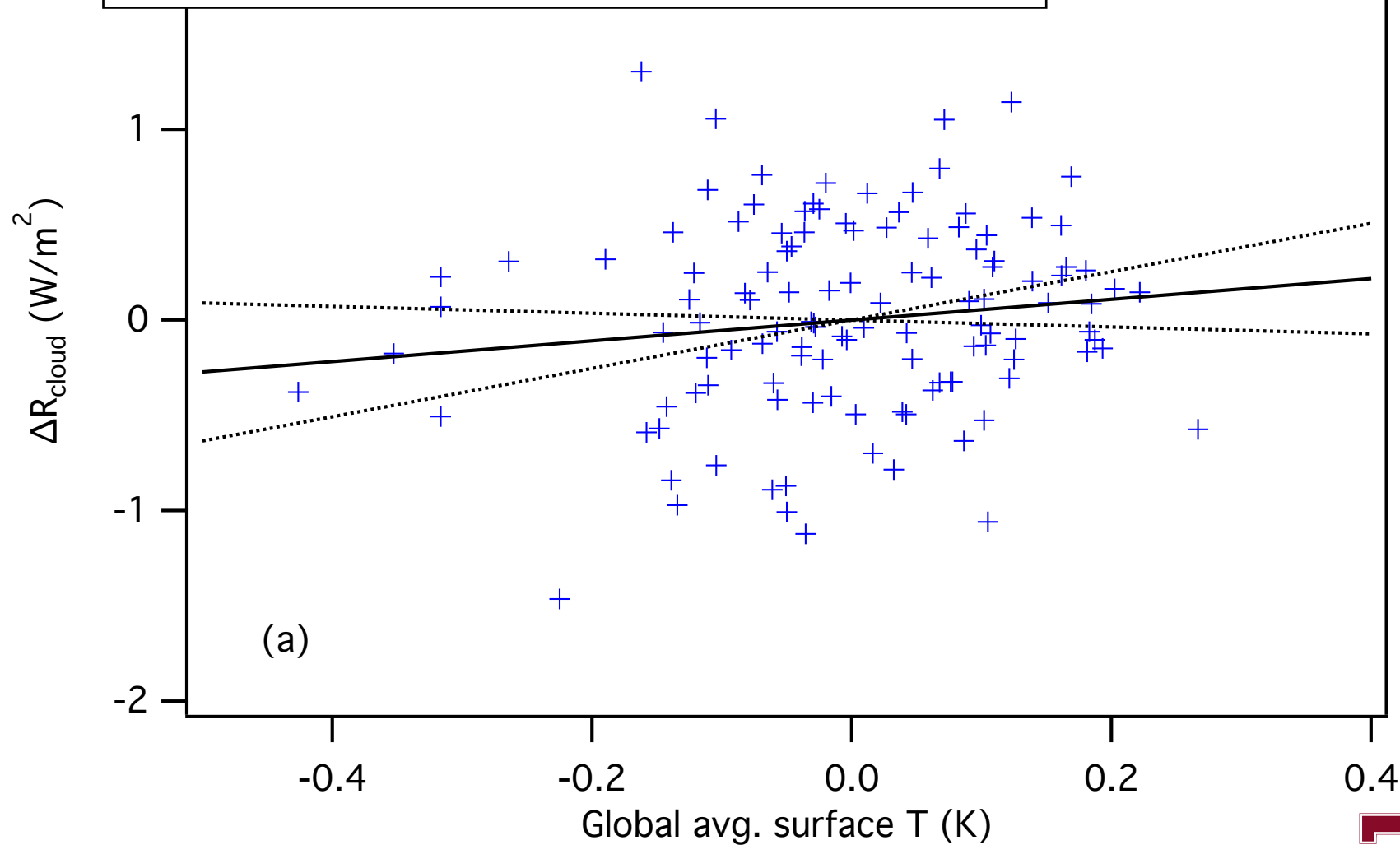


adjustment terms

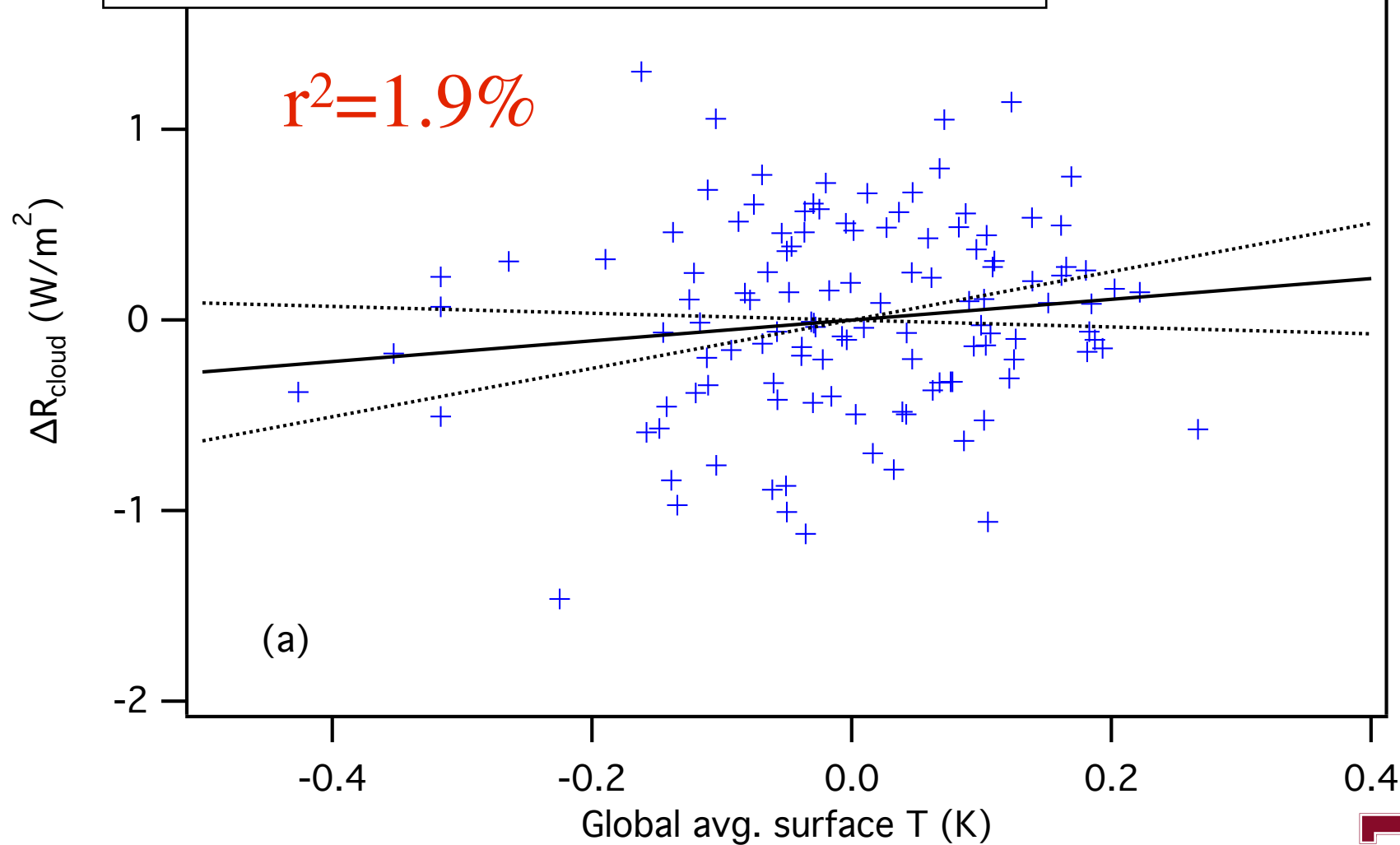




$\lambda_{\text{cloud}} = 0.54 \pm 0.72 \text{ (} 2\sigma \text{) W/m}^2\text{/K (ECMWF)}$   
 $= 0.46 \pm 0.75 \text{ (} 2\sigma \text{) W/m}^2\text{/K (MERRA)}$



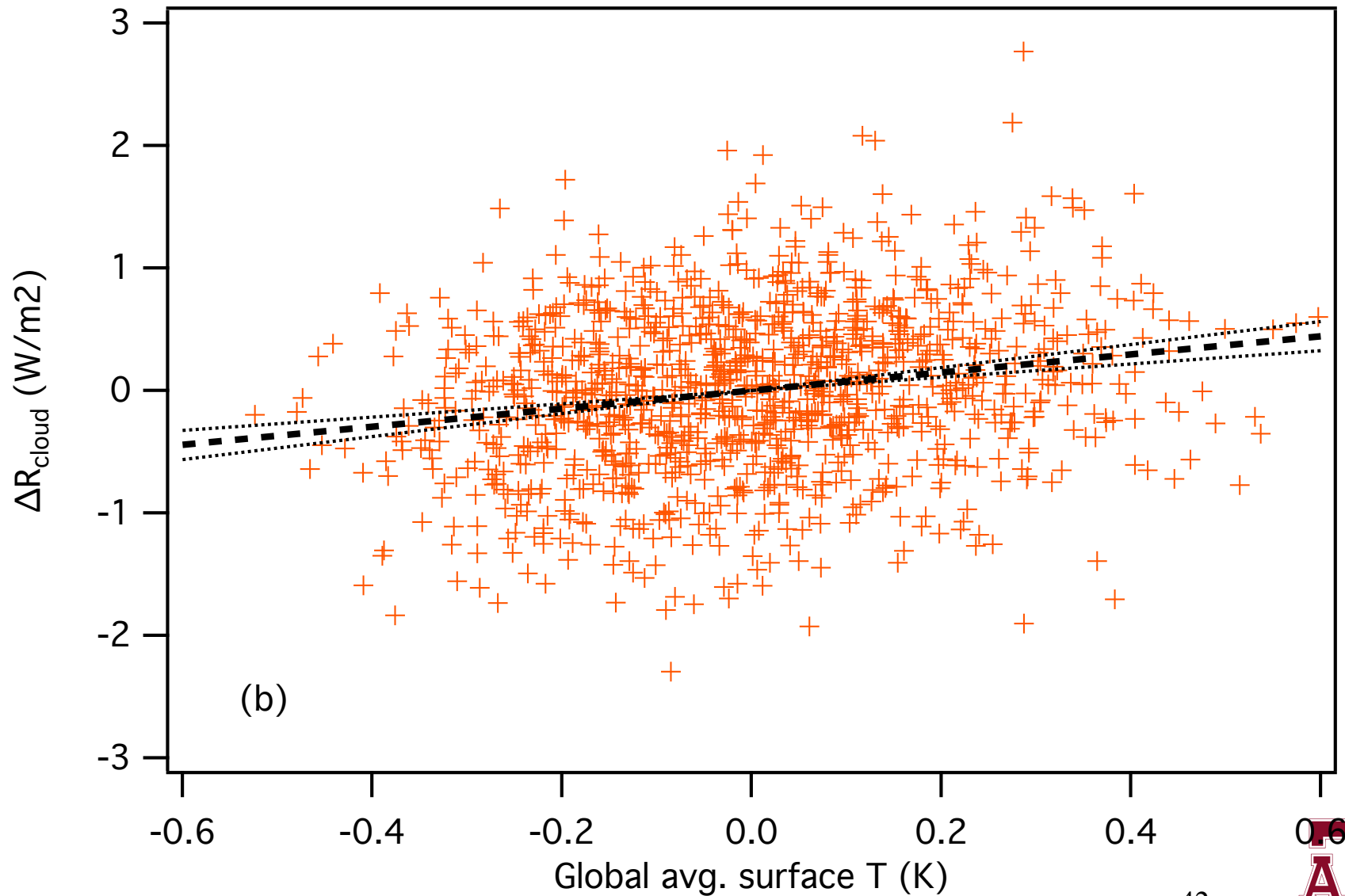
$\lambda_{\text{cloud}} = 0.54 \pm 0.72 \text{ (} 2\sigma \text{) W/m}^2\text{/K (ECMWF)}$   
 $= 0.46 \pm 0.75 \text{ (} 2\sigma \text{) W/m}^2\text{/K (MERRA)}$



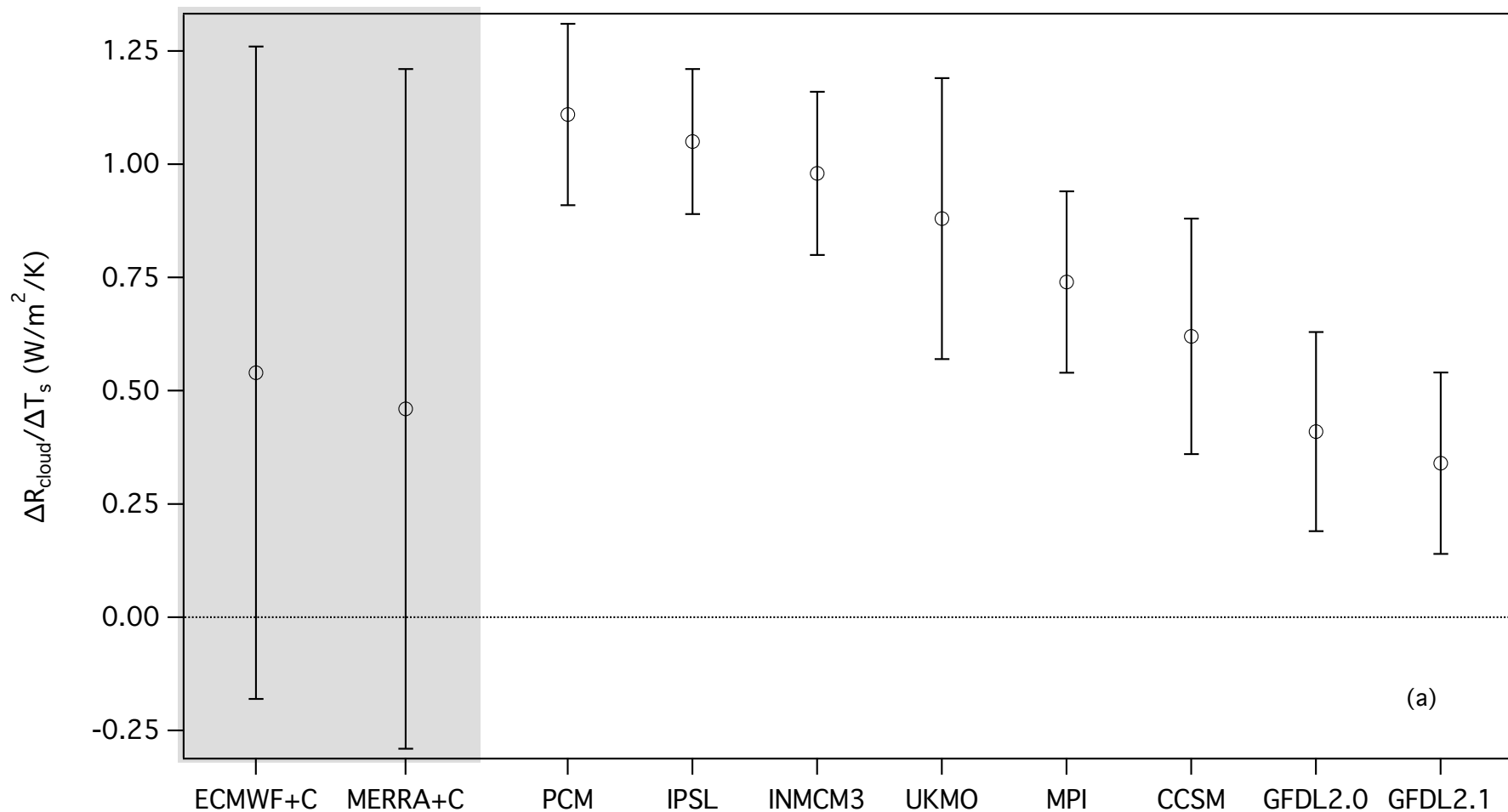
# A few lessons

- This scatter is real
- Another few years of data will not help
- We must study modes of cloud variations that are NOT related to surface T variations
  - e.g., MJO
- Models correctly simulate the scatter





# short-term cloud feedback intercomparison



# Lessons

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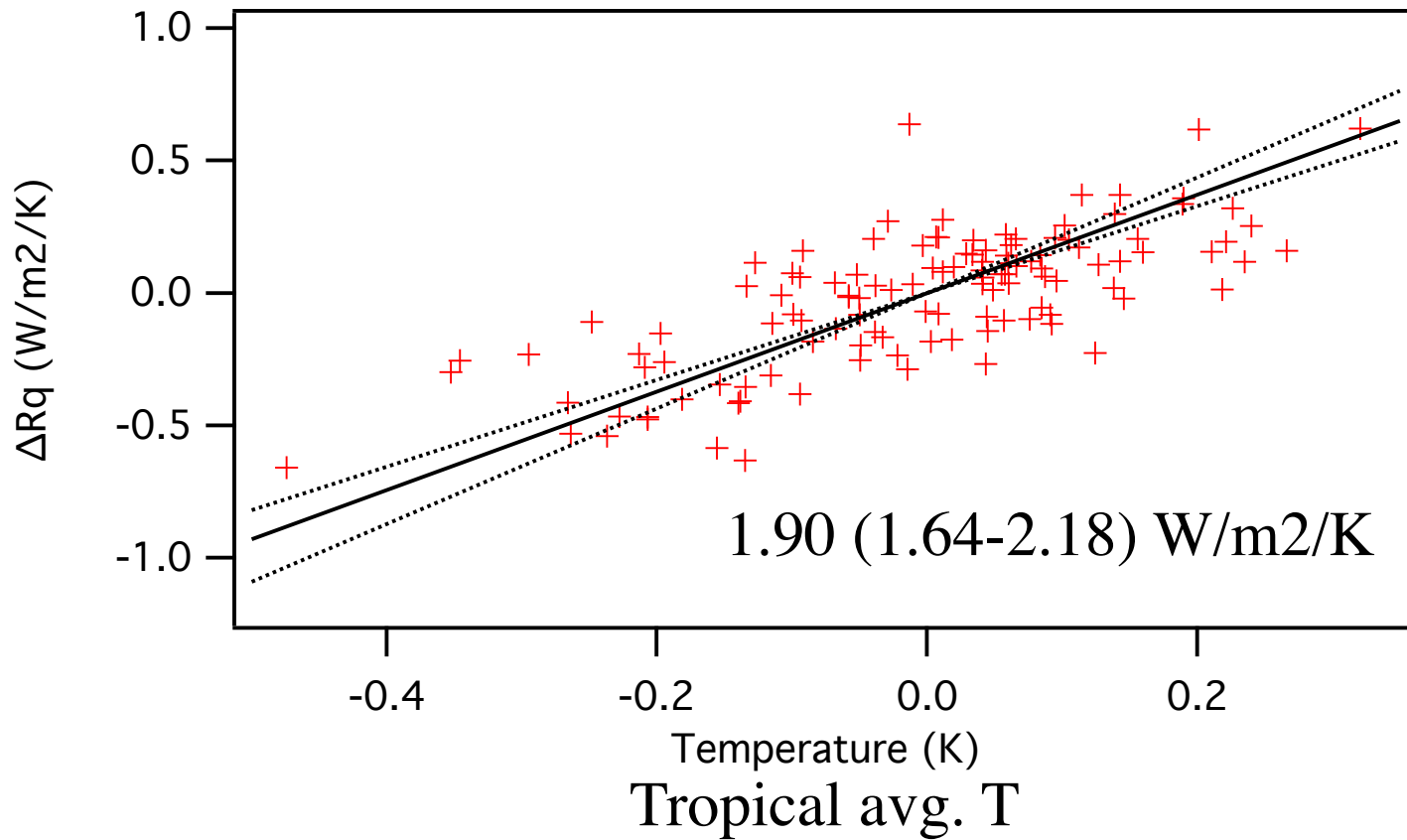
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- This work was supported by NASA grant NNX08AR27G to TAMU

# ECMWF-interim reanalysis

3/2000-2/2010

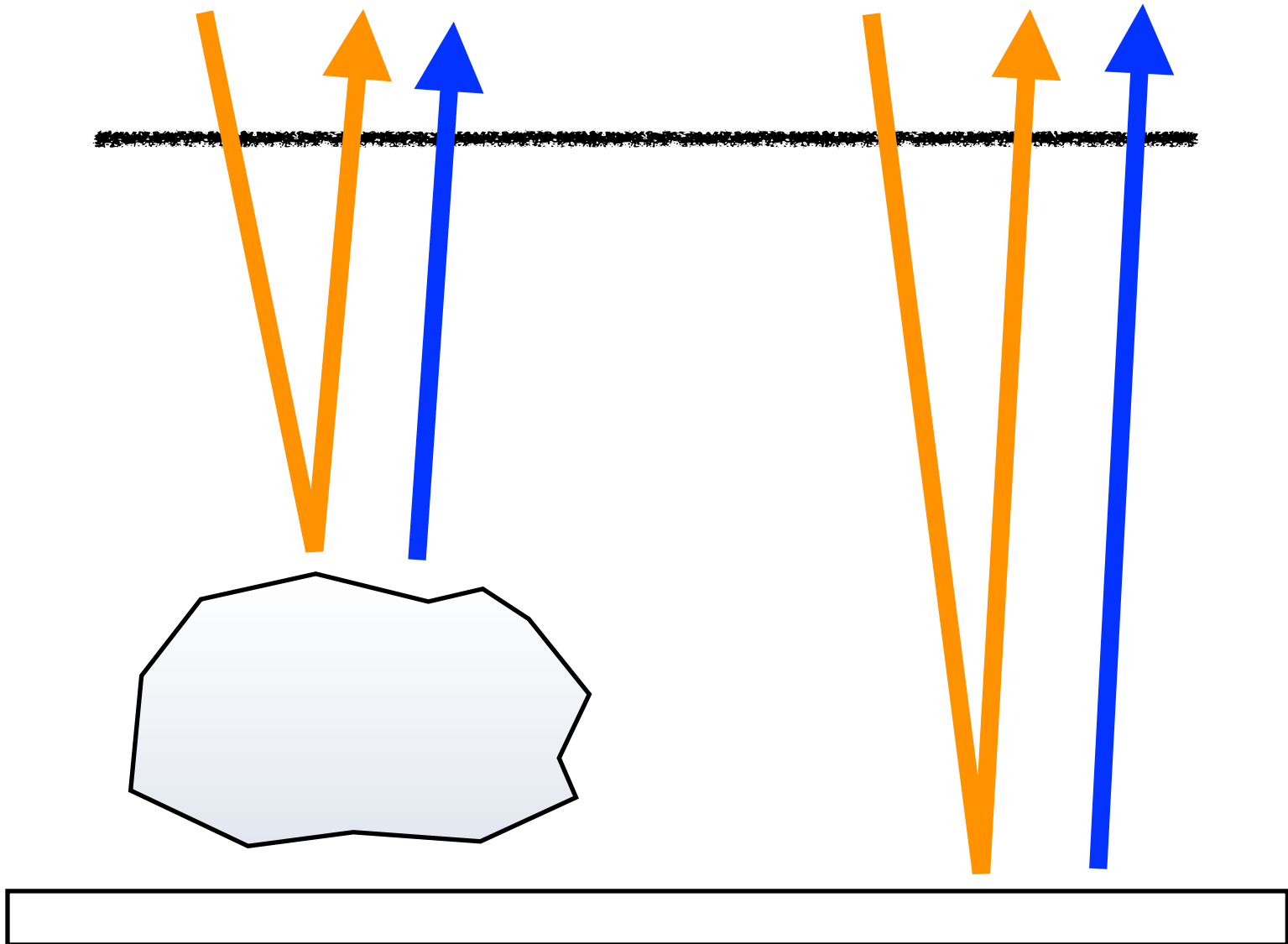




Model	Total		Long wave		Short wave		Long-term cloud feedback	Climate sensitivity
	Cloud feedback	r <sup>2</sup>	Cloud feedback	r <sup>2</sup>	Cloud feedback	r <sup>2</sup>		
FGOALS-g1.0	1.24±0.16	28%	0.92±0.08	48%	0.32±0.15	3%	N/A	2.3
PCM	1.11±0.20	10%	0.52±0.11	7%	0.60±0.21	3%	0.18	2.1
IPSL-CM4	1.05±0.16	12%	1.17±0.13	21%	-0.12±0.14	0.2%	1.06	4.4
INM-CM3.0	0.98±0.18	9%	0.77±0.10	15%	0.21±0.19	0.4%	0.35	2.1
UKMO-HadCM3	0.88±0.31	5%	0.57±0.15	9%	0.31±0.35	0.5%	1.08	3.3
ECHAM/MPI-OM	0.74±0.20	4%	0.97±0.09	27%	-0.23±0.20	0.4%	1.18	3.4
CCSM3	0.62±0.26	2%	0.17±0.12	0.9%	0.45±0.25	1%	0.14	2.7
GFDL-CM2.1	0.34±0.20	0.9%	0.40±0.08	8%	-0.06±0.23	0%	0.81	3.4
GFDL-CM2.0	0.15±0.20	0.2%	-0.63±0.10	11%	0.78±0.21	4%	0.67	2.9
ECMWF-CERES	0.54±0.72	1.9%	0.43±0.45	3.0%	0.12±0.78	0.1%	N/A	N/A
MERRA-CERES	0.46±0.75	1.3%	0.27±0.47	1.2%	0.19±0.76	0.2%	N/A	N/A



$$\text{CRF} = R_{\text{all-sky}} - R_{\text{clear-sky}}$$
$$\text{CRF} = 0$$



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$$\text{CRF} \neq 0$$

